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Pure-Jump Semimartingales

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A new integral with respect to an integer-valued random measure is introduced. In contrast to the finite variation integral ubiquitous in semimartingale theory (Jacod and Shiryaev [6, II.1.5]), the new integral is closed under stochastic integration, composition, and smooth transformations. The new integral gives rise to a previously unstudied class of pure-jump processes — the sigma-locally finite variation pure-jump processes. As an application, it is shown that every semimartingale X has a unique decomposition

$$X = X_0 + X^{\text{qc}} + X^{\text{dp}},$$

where X^{qc} is quasi-left-continuous and X^{dp} is a sigma-locally finite variation pure-jump process that jumps only at predictable times, both starting at zero. The decomposition mirrors the classical result for local martingales (Yoeurp [12, Theoreme 1.4]) and gives a rigorous meaning to the notions of continuous-time and discrete-time components of a semimartingale. Against this backdrop, the paper investigates a wider class of processes that are equal to the sum of their jumps in the semimartingale topology and constructs a taxonomic hierarchy of pure-jump semimartingales.

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1. Introduction

Denote by \mathcal{V}^{d} the set of finite variation pure-jump semimartingales, i.e., those X that are equal to their initial value plus the absolutely convergent sum of their jumps. Equivalently, $X \in \mathcal{V}^{\text{d}}$ if

$$X = X_0 + x * \mu^X,$$

where μ^X is the jump measure of X and $x * \mu^X$ represents the standard jump measure integral (Jacod and Shiryaev [6, II.1.5]). It is known (see [6, II.1.28]) that the predictable compensator of $X \in \mathcal{V}^{\text{d}}$ equals $x * \nu^X$ whenever such compensator exists. Here ν^X denotes the predictable compensator of the jump measure μ^X .

Consider now the \mathbb{R} -valued stochastic process X defined by the following properties:

$$\left. \begin{array}{l} X_0 = 0; \\ X \text{ has independent increments;} \\ \text{jumps of } X \text{ occur only at fixed times } \tau_n = 2^{-1/n}, \text{ for each } n \in \mathbb{N}; \\ \text{the process jumps by } \pm 1/n \text{ with equal probability, for each } n \in \mathbb{N}; \\ X \text{ remains constant outside the fixed jump times.} \end{array} \right\} \quad (1.1)$$

This process is a well-defined semimartingale, in fact a square-integrable martingale, on the whole time line $[0, \infty]$. Yet X is not equal to the sum of its jumps in the conventional sense because the jumps of X are not absolutely summable. In particular, the standard integral $x * \mu^X$ diverges. Furthermore, the integral $x * \nu^X$ also diverges even though the predictable compensator (drift) of X exists and is equal to zero.

The seminal work of Kallsen [7] offers two important clues how to perform symbolic drift calculation for the process in (1.1). The first key idea, [7, Definition 4.1], is to consider each jump time separately and ask only that the expectation $\mathbb{E}[|\Delta X_{\tau_n}|]$ is finite (for each $n \in \mathbb{N}$). One may then be able to legitimately sum up the individual contributions $\mathbb{E}[\Delta X_{\tau_n}]$ over $n \in \mathbb{N}$ even if the sum of $\mathbb{E}[|\Delta X_{\tau_n}|]$ does not converge, which is precisely the case for (1.1). The second key idea, [7, Lemma 4.1], is non-trivial mathematically: it identifies the natural procedure above with the sigma-localization of the absolutely convergent integral $x * \nu^X$ to obtain a better integral that we shall denote $x \star \nu^X$. It is important to add that Kallsen [7] makes everything work in a general continuous-time setting that goes well beyond (1.1).

Acting on these clues, we propose to view the process (1.1) in two novel, complementary ways, both of which involve an approximation by elements in \mathcal{V}^d . The first approach regards X in (1.1) as an element of \mathcal{V}_σ^d , i.e., as a process that belongs sigma-locally to \mathcal{V}^d . Proposition 3.12 shows that each $X \in \mathcal{V}_\sigma^d$ satisfies the convenient formula

$$X = X_0 + x \star \mu^X,$$

where \star is the sigma-localized version of the standard jump measure integral $*$. Furthermore, the drift of X , provided it exists, is given by $x \star \nu^X$ (Corollary 3.13).

The new jump integral \star , unlike $*$, is associative so that for a predictable process ζ and a predictable function η one has

$$\zeta \cdot (\eta \star \mu^X) = (\zeta \eta) \star \mu^X$$

if the left-hand side is well defined (Proposition 3.9). The associativity of the stochastic integral is in fact a special case of a stronger composition property

$$\psi \star (\eta \star \mu^X) = \psi(\eta) \star \mu^X,$$

which holds as soon as one of the two expressions is meaningful (Proposition 3.8). The associativity of integral then follows by taking for ψ the linear function $\psi(x) = \zeta x$. Together with closedness under \mathcal{C}^2 transformations (Proposition 3.10) these properties give rise to a very pleasant stochastic calculus for processes in \mathcal{V}_σ^d . For further developments in this direction, see Černý and Ruf [1].

The second approach views X as a sum of its jumps at a sequence of stopping times with convergence in the Émery semimartingale topology. Here it is in principle possible to encounter two different processes that share the same jump measure (by choosing different exhausting sequence of stopping times in each case). However, we show that such a situation cannot occur in \mathcal{V}_σ^d (Theorem 4.3). This offers a wider sense in which all processes in \mathcal{V}_σ^d are uniquely determined by their jump measure. In contrast, the family of quadratic pure-jump processes (Protter [11, p. 63]; see also Definition 4.1) lacks such uniqueness property as adding a continuous deterministic finite variation process to X yields a different quadratic pure-jump process with the same jump measure. Thus the terminology “quadratic pure-jump” is unfortunate — in the context of this paper, most of these processes are not pure-jump processes at all. In fact, it is shown that a special quadratic pure-jump semimartingale is in \mathcal{V}_σ^d if and only if its drift is given by $x \star \nu^X$ (Corollary 4.4).

The process (1.1) has one property that not all processes in \mathcal{V}_σ^d share: its jumps are exhausted by a sequence of *predictable* (in fact fixed) times. Such processes enjoy further regularity in that *any* sequence of predictable times exhausting their jumps delivers convergence of the sum in the Émery topology (Proposition 4.6). As an application, it is shown that every semimartingale X has a unique decomposition

$$X = X_0 + X^{\text{qc}} + X^{\text{dp}},$$

where X^{qc} is quasi-left-continuous and X^{dp} is an element of \mathcal{V}_σ^d that jumps only at predictable times, both starting at zero (Proposition 3.15). The decomposition mirrors the classical result for local martingales (Yoeurp [12, Theoreme 1.4]) and gives a rigorous meaning to the notions of continuous-time and discrete-time components of a semimartingale. The decomposition is helpful when computing drifts in various applications; for further details we refer the reader to Černý and Ruf [1, Section 6].

Here now the outline of this paper. After notation (Section 2), we develop the new integral (Section 3) and then proceed with the classification of the pure-jump processes (Section 4). Section 5 has proofs pertaining to Section 4. Appendix A contains facts about the Émery topology and Appendix B discusses the consequences of weakening the topology to the one given by uniform convergence in probability.

2. Notation and setup

We fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ with a right-continuous filtration $\mathfrak{F} = (\mathcal{F}_t)_{t \geq 0}$. Recall from [6, II.1.4] that a function $\eta : \Omega \times [0, \infty) \times \mathbb{R} \rightarrow \mathbb{R}$ is called predictable if it is $\mathcal{P} \times \mathcal{B}(\mathbb{R})$ -measurable, where \mathcal{P} denotes the predictable sigma field and $\mathcal{B}(\mathbb{R})$ the Borel sigma field on \mathbb{R} . If $\psi : \Omega \times [0, \infty) \times \mathbb{R} \rightarrow \mathbb{R}$ denotes another (predictable) function we shall write $\psi(\eta)$ to denote the (predictable) function $(\omega, t, x) \mapsto \psi(\omega, t, \eta(\omega, t, x))$.

We shall consider an integer-valued random measure μ on $[0, \infty) \times \mathbb{R}$ with predictable compensator ν . A predictable function η with $\eta(0) = 0$ is integrable with respect to μ , i.e., $\eta * \mu$ exists if $|\eta| * \mu < \infty$. Recall from [6, II.2.9] that ν can be written in disintegrated form as

$$\nu(dt, dx) = F_t(dx) dA_t, \quad t \geq 0, x \in \mathbb{R}, \quad (2.1)$$

where A is a nondecreasing predictable process, and F is a transition kernel from $(\Omega \times [0, \infty), \mathcal{P})$ into $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$. If we want to emphasize the probability measure under which ν is the predictable compensator of μ we shall write $\nu(\mathbb{P})$.

We let \mathcal{S} denote the space of \mathbb{R} -valued semimartingales. For a semimartingale $X \in \mathcal{S}$, we let X_- denote its left-limit process with the convention $X_{0-} = X_0$ and we let $\Delta X = X - X_-$ denote its jump process. Next, we let μ^X denote the jump measure of X and ν^X its predictable compensator. For a predictable function η with $\eta(0) = 0$ we then have $\eta * \mu^X = \sum_{0 < t \leq \cdot} \eta(\Delta X_t)$ if $|\eta| * \mu^X < \infty$. The corresponding quantities in (2.1) shall be written with a superscript X . If X is special, we write B^X for its drift, i.e., B^X is the predictable finite variation process with $B_0^X = 0$ such that $X - B^X$ is a local martingale. If Y denotes another semimartingale then $[X, Y]$ denotes the quadratic covariation of X and Y . Moreover, we write $X[1] = X - x \mathbf{1}_{|x| > 1} * \mu^X$ and note that $X[1]$ is special. Next, $L(X)$ denotes the family of X -integrable predictable processes.

If $\mathcal{J} \subset \mathcal{S}$ is a family of semimartingales, we say that a semimartingale Y belongs to \mathcal{J}_σ , the sigma-localized class of \mathcal{J} , if there is a sequence $(D_n)_{n \in \mathbb{N}}$ of predictable sets increasing

to $\Omega \times [0, \infty)$ such that $\mathbf{1}_{D_n} \cdot Y \in \mathcal{J}$ for each $n \in \mathbb{N}$. We say that \mathcal{J} is stable under sigma-stopping (see Kallsen [7, Definition 2.1]) if for every $X \in \mathcal{J}$ and every predictable set D the process $\mathbf{1}_D \cdot X$ belongs to \mathcal{J} . Finally, we shall say that \mathbf{Q} is a probability measure that is locally absolutely continuous with respect to \mathbf{P} if \mathbf{Q} is absolutely continuous with respect to \mathbf{P} on \mathcal{F}_t for each $t \geq 0$.

Remark 2.1. Throughout this paper, we only consider the scalar case, which helps in reducing notation. The careful reader can convince themselves that quite a few results (in particular those of Section 3) generalize to the higher-dimensional case, for example when X takes values in \mathbb{R}^d and the predictable functions below map into \mathbb{R}^n , etc., for some $d, n \in \mathbb{N}$. A notable exception is statement (ii) in Theorem 4.3, where we do not know whether the one-dimensional situation generalizes. Indeed, we do not know whether \mathcal{J}^2 is a vector space — the lack of such structure would seem to imply that such a result does not hold in higher dimensions. \square

3. Extended integral with respect to a random measure

We start by extending the standard definition of integral with respect to a random measure and derive some basic properties in Subsection 3.1. Then, in Subsection 3.2, we prove some associativity properties of this integral. In Subsection 3.3 we connect the integral to the representation of sigma-locally finite variation pure-jump processes. In particular, this will enable us to write the process X of the introduction as $X = x \star \mu^X$ and its drift under any locally absolutely continuous measure \mathbf{Q} that makes X special as $B^X(\mathbf{Q}) = x \star \nu^X(\mathbf{Q})$.

3.1. Definition and basic properties of the extended integral

Definition 3.1 (Extended integral with respect to random measure).

- (i) Denote by $L(\mu)$ the set of predictable functions that are absolutely integrable with respect to μ . We say that a predictable function η belongs to $L_\sigma(\mu)$, the sigma-localized class of $L(\mu)$, if there is a sequence $(D_n)_{n \in \mathbb{N}}$ of predictable sets increasing to $\Omega \times [0, \infty)$ and a semimartingale Y such that $\mathbf{1}_{D_n} \eta \in L(\mu)$ for each $n \in \mathbb{N}$ and

$$(\mathbf{1}_{D_n} \eta) * \mu = \mathbf{1}_{D_n} \cdot Y, \quad n \in \mathbb{N}.$$

In such case the semimartingale Y is denoted by $\eta \star \mu$.

- (ii) $L(\nu)$, $L_\sigma(\nu)$, and $\eta \star \nu$ are defined analogously; see Definition 4.1 and Lemma 4.1 in Kallsen [7]. \square

Note that if $\mu = \nu$ is a predictable random measure then the two definitions above agree; hence $L_\sigma(\mu)$ and $L_\sigma(\nu)$ are well-defined and we have $L_\sigma(\mu) = L_\sigma(\nu)$. Note also that $\eta \star \mu$ (resp., $\eta \star \nu$) is uniquely defined provided that $\eta \in L_\sigma(\mu)$ (resp., $\eta \in L_\sigma(\nu)$).

Remark 3.2. Let \mathbf{Q} denote a probability measure locally absolutely continuous with respect to \mathbf{P} . With the obvious notation, we then have $L_\sigma^\mathbf{P}(\mu) \subset L_\sigma^\mathbf{Q}(\mu)$. For $L_\sigma^\mathbf{P}(\nu(\mathbf{P}))$ and $L_\sigma^\mathbf{Q}(\nu(\mathbf{Q}))$, no such inclusions hold in general. However, refer also to the positive statement in Remark 3.5. \square

The following characterization of $L_\sigma(\nu)$ appears in the literature.

Lemma 3.3 (Kallsen [7, Definition 4.1 and Lemma 4.1]). *For a predictable function η the following statements are equivalent.*

- (i) $\eta \in L_\sigma(\nu)$.
- (ii) *The following two conditions hold.*
 - (a) $\int |\eta_t(x)| F_t(dx) < \infty$ $(\mathbf{P} \times dA)$ -a.e.
 - (b) $\int \left| \int \eta_t(x) F_t(dx) \right| dA_t < \infty$.

Moreover, for $\eta \in L_\sigma(\nu)$ one has

$$\eta \star \nu = \int_0^\cdot \left(\int \eta_t(x) F_t(dx) \right) dA_t.$$

To the best of our knowledge, the class $L_\sigma(\mu)$ has not been studied previously. The following characterization therefore seems to be new.

Proposition 3.4. *For a predictable function η the following statements are equivalent.*

- (i) $\eta \in L_\sigma(\mu)$.
- (ii) *The following two conditions hold.*
 - (a) $\eta^2 \star \mu < \infty$.
 - (b) $\eta \mathbf{1}_{\{|\eta| \leq 1\}} \in L_\sigma(\nu)$.

Furthermore, for $\eta \in L_\sigma(\mu)$ one has

$$\eta \star \mu = \eta \mathbf{1}_{\{|\eta| > 1\}} \star \mu + \eta \mathbf{1}_{\{|\eta| \leq 1\}} \star (\mu - \nu) + \eta \mathbf{1}_{\{|\eta| \leq 1\}} \star \nu, \quad (3.1)$$

where the integral with respect to the compensated measure $\mu - \nu$ is defined in [6, II.1.27(b)].

Remark 3.5. In the setup of Remark 3.2, choose a predictable function η with $|\eta|^2 \star \mu < \infty$. Proposition 3.4 now yields that if $\eta \mathbf{1}_{\{|\eta| \leq 1\}} \in L_\sigma^P(\nu(\mathbf{P}))$ then also $\eta \mathbf{1}_{\{|\eta| \leq 1\}} \in L_\sigma^Q(\nu(\mathbf{Q}))$. \square

Proof of Proposition 3.4. In the following we argue both inclusions and (3.1).

(i) \Rightarrow (ii): Let $(D_n)_{n \in \mathbb{N}}$ be as in Definition 3.1(i). Then $\mathbf{1}_{D_n} |\eta|^2 \star \mu = \mathbf{1}_{D_n} \cdot [\eta \star \mu, \eta \star \mu]$ for all $n \in \mathbb{N}$, and a monotone convergence argument yields $|\eta|^2 \star \mu = [\eta \star \mu, \eta \star \mu] < \infty$. Let us now set $\bar{\eta} = \eta \mathbf{1}_{\{|\eta| \leq 1\}}$. Then $\bar{\eta} \in L_\sigma(\mu)$ and we directly get

$$\int_0^\cdot \mathbf{1}_{D_n}(t) \left(\int |\bar{\eta}_t(x)| F_t(dx) \right) dA_t = \mathbf{1}_{D_n} |\bar{\eta}| \star \nu < \infty.$$

Thanks to Lemma 3.3 we now only need to argue that $\int \left| \int \bar{\eta}_t(x) F_t(dx) \right| dA_t < \infty$. We note that $|\Delta(\bar{\eta} \star \mu)| \leq 1$, hence $\bar{\eta} \star \mu$ is special, say with predictable finite variation drift \bar{B} . By monotone convergence, we now get

$$\begin{aligned} \int_0^\cdot \left| \int \bar{\eta}_t(x) F_t(dx) \right| dA_t &= \lim_{n \uparrow \infty} \int_0^\cdot \mathbf{1}_{D_n}(t) \left| \int \bar{\eta}_t(x) F_t(dx) \right| dA_t = \lim_{n \uparrow \infty} \int_0^\cdot \mathbf{1}_{D_n}(t) |d\bar{B}_t| \\ &= \int_0^\cdot |d\bar{B}_t| < \infty. \end{aligned}$$

This yields $\bar{\eta} \in L_\sigma(\nu)$, hence the implication (i) \Rightarrow (ii) is shown.

(ii) \Rightarrow (i) and (3.1): Let $(D_n)_{n \in \mathbb{N}}$ be as in Definition 3.1(ii). Note that all terms on the right-hand side of (3.1) are well defined and yield a semimartingale Y provided that (ii) holds. Thanks to the uniqueness of $\eta \star \mu$ we only need to observe that $\mathbf{1}_{D_n} \cdot Y = (\mathbf{1}_{D_n} \eta) \star \mu$ for all $n \in \mathbb{N}$. However, this is straightforward, which concludes the proof of the proposition. \square

Remark 3.6. Note that $L_{\text{loc}}(\mu) = L(\mu)$, that is $L(\mu)$ is closed under standard localization. However, we have $L_\sigma(\mu) \supsetneq L(\mu)$ on sufficiently large probability spaces; see Example 3.7. \square

Example 3.7. Let μ denote a jump measure with $\nu(dt, dx) = \frac{1}{2} \mathbf{1}_{|x| \leq 1} dx (\sum_{i=1}^\infty \delta_{1/i}(dt))$, where $\delta_{1/i}$ denotes the Dirac measure at $1/i$. Then for the decomposition in (2.1) we may choose $A = \sum_{i=1}^\infty \frac{1}{i^2} \mathbf{1}_{[1/i, \infty[}$ and $F_t(dx) = \frac{1}{(2t^2)} \mathbf{1}_{|x| \leq 1} dx$ for all $t > 0$ and $x \in \mathbb{R}$.

Consider next the predictable function η given by $\eta_t(x) = tx$ for all $t \geq 0$ and $x \in \mathbb{R}$. Then $\eta \in L_\sigma(\nu)$ by Lemma 3.3 and $\eta^2 \star \mu_\infty \leq \sum_{i=1}^\infty \frac{1}{i^2} < \infty$. Hence by Proposition 3.4, we have $\eta \in L_\sigma(\mu)$. Indeed, $\eta \star \mu$ is a semimartingale that jumps at times $1/i$ and is constant on the intervals $[1/(i+1), 1/i]$, for each $i \in \mathbb{N}$. Moreover, $\Delta(\eta \star \mu)_{1/i} = U_i/i$, where $(U_i)_{i \in \mathbb{N}}$ is a sequence of independent $[-1, 1]$ -uniforms. Since $|\eta| \star \mu_1 = \sum_{i=1}^\infty |U_i|/i = \infty$ (by Kolmogorov's convergence criteria), we have $\eta \in L_\sigma(\mu) \setminus L(\mu)$.

Consider now the predictable function $\bar{\eta}_t(x) = x$ for all $t \geq 0$ and $x \in \mathbb{R}$. Then again $\bar{\eta} \in L_\sigma(\nu)$, but now $\bar{\eta}^2 \star \mu_1 = \infty$. This yields an example for a predictable function $\bar{\eta} \in L_\sigma(\nu) \setminus L_\sigma(\mu)$. \square

3.2. Associativity properties of the extended integral

We remind the reader that μ without a superscript refers to a given integer-valued random measure, while μ^X refers to the jump measure of a semimartingale X ; see Section 2.

Proposition 3.8. Let $\eta \in L_\sigma(\mu)$ and $\psi : \Omega \times [0, \infty) \times \mathbb{R} \rightarrow \mathbb{R}$ be a predictable function. Then the following statements are equivalent.

- (i) $\psi \in L_\sigma(\mu^{\eta \star \mu})$.
- (ii) $\psi(\eta) \in L_\sigma(\mu)$.

Furthermore, if either condition holds then $\psi \star (\eta \star \mu) = \psi(\eta) \star \mu$. Moreover, the same assertions hold with μ replaced by ν .

Proof. Let us first prove the statement with μ replaced by ν . To this end, note that

$$\nu^{\eta \star \nu}(dt, dx) = \bar{F}_t(dx) dA_t, \quad t \geq 0, x \in \mathbb{R},$$

where \bar{F}_t is the image of measure F_t under η_t . Then the equivalence follows from Lemma 3.3. The statement for μ follows exactly in the same manner, now using Proposition 3.4. \square

Next, we prove a composition property for stochastic integrals. Recall that $L(\eta \star \mu)$ denotes the set of predictable processes that are integrable with respect to the semimartingale $\eta \star \mu$.

Proposition 3.9. Let $\eta \in L_\sigma(\mu)$ and $\zeta : \Omega \times [0, \infty) \rightarrow \mathbb{R}$ be a predictable process. Then the following statements are equivalent.

- (i) $\zeta \in L(\eta \star \mu)$.

(ii) $\zeta\eta \in L_\sigma(\mu)$.

Furthermore, if either condition holds then $\zeta \cdot (\eta \star \mu) = (\zeta\eta) \star \mu$. Moreover, the same assertions hold with μ replaced by ν .

Proof. We shall prove the statement only for μ as the same argument works if μ is replaced by ν . Note that there is a sigma-localizing sequence $(D_n)_{n \in \mathbb{N}}$ such that

$$D_n \subset \{(\omega, t) \in \Omega \times [0, \infty) : |\zeta(\omega, t)| \leq n\}$$

and $\mathbf{1}_{D_n}\eta \in L(\mu)$ with $\mathbf{1}_{D_n} \cdot (\eta \star \mu) = (\mathbf{1}_{D_n}\eta) \star \mu$. Note that for each $n \in \mathbb{N}$ we have $\zeta\mathbf{1}_{D_n} \in L(\eta \star \mu)$ and $\zeta\mathbf{1}_{D_n}\eta \in L(\mu)$. Then if (i) holds we have

$$\mathbf{1}_{D_n} \cdot (\zeta \cdot (\eta \star \mu)) = (\zeta\mathbf{1}_{D_n}) \cdot (\mathbf{1}_{D_n} \cdot (\eta \star \mu)) = (\zeta\mathbf{1}_{D_n}) \cdot (\mathbf{1}_{D_n}\eta \star \mu) = (\zeta\mathbf{1}_{D_n}\eta) \star \mu,$$

for each $n \in \mathbb{N}$. This yields (ii) and $\zeta \cdot (\eta \star \mu) = (\zeta\eta) \star \mu$.

Assume now that (ii) holds. We then have

$$\begin{aligned} \mathbf{1}_{D_n} \cdot ((\zeta\eta) \star \mu) &= (\mathbf{1}_{D_n}\zeta\eta) \star \mu = (\zeta\mathbf{1}_{D_n}) \cdot (\mathbf{1}_{D_n}\eta \star \mu) \\ &= (\zeta\mathbf{1}_{D_n}) \cdot (\mathbf{1}_{D_n} \cdot (\eta \star \mu)) = (\zeta\mathbf{1}_{D_n}) \cdot (\eta \star \mu) \end{aligned}$$

for each $n \in \mathbb{N}$. An application of [7, Lemma 2.2] then yields that (i) holds. \square

Proposition 3.10. *Let $\eta \in L_\sigma(\mu)$ and $f : \mathbb{R} \rightarrow \mathbb{R}$ be a twice continuously differentiable function. Then for $Y = Y_0 + \eta \star \mu$ we have $\xi = f(Y_- + \eta) - f(Y_-) \in L_\sigma(\mu)$ and $f(Y) = f(Y_0) + \xi \star \mu$.*

Proof. For each $n \in \mathbb{N}$ let τ_n denote the first time that $|Y|$ is greater than or equal to n . Recall that $f(Y)$ is a semimartingale and that there exists a sequence $(D_n)_{n \in \mathbb{N}}$ of predictable sets increasing to $\Omega \times [0, \infty)$ such that $\mathbf{1}_{D_n}\eta \in L(\mu)$. Without loss of generality we may assume that $D_n \subset [0, \tau_n]$ for each $n \in \mathbb{N}$.

It suffices now to argue that $\mathbf{1}_{D_n}\xi \in L(\mu)$ and that $\mathbf{1}_{D_n} \cdot f(Y) = (\mathbf{1}_{D_n}\xi) \star \mu$ for each $n \in \mathbb{N}$. To argue these two claims, fix $n \in \mathbb{N}$. Observe now that

$$\mathbf{1}_{D_n}|\xi| \leq \mathbf{1}_{D_n}\mathbf{1}_{\{|\eta| > 1\}}|\xi| + L_{n+1}\mathbf{1}_{D_n}\mathbf{1}_{\{|\eta| \leq 1\}}|\eta|,$$

where L_{n+1} is the Lipschitz constant of f restricted to the domain $[-n-1, n+1]$. The first term on the right-hand side of the last display is in $L(\mu)$ as it only contributes finitely many terms and so is the second term as it is bounded by a constant times $\mathbf{1}_{D_n}|\eta|$. Hence we have argued the first claim, namely that $\mathbf{1}_{D_n}\xi \in L(\mu)$. Next, note that Itô's formula and Proposition 3.8 yield

$$\begin{aligned} \mathbf{1}_{D_n} \cdot f(Y) &= \mathbf{1}_{D_n} \cdot (f'(Y_-) \cdot Y + (\xi - f'(Y_-)\eta) \star \mu) \\ &= f'(Y_-) \cdot (\mathbf{1}_{D_n} \cdot Y) + (\mathbf{1}_{D_n}\xi - \mathbf{1}_{D_n}f'(Y_-)\eta) \star \mu = (\mathbf{1}_{D_n}\xi) \star \mu. \end{aligned}$$

This is the second claim, and the assertion follows. \square

Example 3.11. *As a counterpoint to Proposition 3.9, we will now exhibit an integer-valued random measure μ with the following properties.*

(1) $x \in L_\sigma(\mu)$.

- (2) $x \star \mu$ is quasi-left-continuous with bounded jumps.
- (3) there is a predictable process $\zeta \in L(x \star (\mu - \nu))$ such that $\zeta \notin L(x \star \mu)$.

To this end, let N denote a standard Poisson process. That is, N jumps up by one with standard exponentially distributed waiting times and $B_t^N = t$ for all $t \geq 0$. Let now $\varphi_t = 1/k$ for all $t \in [k-1, k)$ and all $k \in \mathbb{N}$, and fix $n \in \{1, 2\}$. Then $\varphi^n \in L(N) \cap L(B^N)$ and $\varphi^n \cdot N = \varphi^n \cdot (N - B^N) + \varphi^n \cdot B^N$ is the sum of a uniformly integrable martingale and an increasing process (of bounded variation in the case $n = 2$). Indeed, Kolmogorov's two-series theorem, applied to the sequence $(\varphi^n \cdot (N - B^N)_k)_{k \in \mathbb{N}}$, and an application of the Borel-Cantelli lemma, or Larsson and Ruf [9, Corollary 4.4], yield the existence of the random variable $\varphi^n \cdot (N - B^N)_\infty = \lim_{t \uparrow \infty} \varphi^n \cdot (N - B^N)_t$. The Burkholder-Davis-Gundy inequality yields that $\varphi^n \cdot (N - B^N)$ is a uniformly integrable martingale as claimed.

In particular, it follows that the process $Y_t = \varphi^2 \cdot N_{\tan(t \wedge \pi/2)}$ is a special semimartingale on the whole time line with

$$B_t^Y = \int_0^{\tan(t \wedge \pi/2)} \varphi_u^2 du, \quad t \geq 0.$$

Statements (1)–(3) now follow by taking $\mu = \mu^Y$, $\zeta_t = 1/\varphi_{\tan(t)} \mathbf{1}_{t < \pi/2}$ and observing that $x \star \mu = Y$ and

$$\zeta \cdot (x \star (\mu - \nu))_t = \zeta \cdot (Y - B^Y)_t = \varphi \cdot (N - B^N)_{\tan(t \wedge \pi/2)}, \quad t \geq 0.$$

From $\lim_{t \uparrow \infty} \varphi \cdot B_t^N = \infty$ we obtain $\zeta \notin L(B^Y)$, whereby $\zeta \in L(Y - B^Y)$ yields $\zeta \notin L(Y) = L(x \star \mu)$. \square

3.3. Sigma-locally finite variation pure-jump processes and the extended integral

The statements in the previous subsections can also be expressed in terms of the class \mathcal{V}_σ^d .

Proposition 3.12. *If $\eta \in L_\sigma(\mu)$ then $\eta \star \mu \in \mathcal{V}_\sigma^d$. Conversely, if $X \in \mathcal{V}_\sigma^d$ then $x \in L_\sigma(\mu^X)$ and*

$$\begin{aligned} X &= X_0 + x \star \mu^X \\ &= X_0 + x \mathbf{1}_{|x| > 1} \star \mu^X + x \mathbf{1}_{|x| \leq 1} \star (\mu^X - \nu^X) + x \mathbf{1}_{|x| \leq 1} \star \nu^X. \end{aligned} \quad (3.2)$$

Proof. The first part of the assertion follows directly from the definitions of $L_\sigma(\mu^X)$ and \mathcal{V}_σ^d . The second equality in (3.2) is the consequence of Proposition 3.4. \square

Corollary 3.13. *Let $X \in \mathcal{V}_\sigma^d$. Then the following statements are equivalent.*

- (i) X is special.
- (ii) $x \in L_\sigma(\nu)$.

Furthermore, if either condition holds then $B^X = x \star \nu^X$.

Proof. By [6, Lemma I.4.24], the semimartingale X is special if and only if $x \mathbf{1}_{|x| > 1} \star \mu^X$ is special. By [6, II.1.28], the finite variation pure-jump semimartingale $x \mathbf{1}_{|x| > 1} \star \mu^X$ is special if and only if $x \mathbf{1}_{|x| > 1} \in L(\nu^X)$, in which case the drift of $x \mathbf{1}_{|x| > 1} \star \mu^X$ equals $x \mathbf{1}_{|x| > 1} \star \nu^X$. The claim now follows from Proposition 3.12 and [6, Definition II.1.27]. \square

Corollary 3.14. *Let $X \in \mathcal{V}_\sigma^d$, $\zeta \in L(X)$, and f be a twice continuously differentiable function. Then $\zeta \cdot X, f(X) \in \mathcal{V}_\sigma^d$ with*

$$\begin{aligned}\zeta \cdot X &= (\zeta x) \star \mu^X; \\ f(X) &= f(X_0) + (f(X_- + x) - f(X_-)) \star \mu^X.\end{aligned}$$

Proof. This follows from Proposition 3.12 in conjunction with Propositions 3.9 and 3.10. \square

Recall that semimartingale X is said to be quasi-left-continuous if $\Delta X_\tau = 0$ almost surely on $\{\tau < \infty\}$ for each predictable time τ . Yoeurp [12] has shown that every local martingale can be uniquely decomposed into two components, one quasi-left-continuous and the other with jumps only at predictable times, such that the quadratic covariation of the two components is zero. This motivates the following result.

Proposition 3.15. *Every semimartingale X has the unique decomposition*

$$X = X_0 + X^{\text{qc}} + X^{\text{dp}}, \quad (3.3)$$

where $X_0^{\text{qc}} = X_0^{\text{dp}} = 0$, X^{qc} is a quasi-left-continuous semimartingale, X^{dp} jumps only at predictable times, and $X^{\text{dp}} \in \mathcal{V}_\sigma^d$. We then have $[X^{\text{qc}}, X^{\text{dp}}] = 0$.

Proof. Let τ denote any predictable time. Note that $\Delta X_\tau = \Delta X_\tau^{\text{dp}}$ for any decomposition of X by the quasi-left-continuity of X^{qc} . This proves the uniqueness of the decomposition. Consider now the predictable process $(x^2 \wedge 1) \star \nu^X$. Applying [6, I.2.24] yields a family $(\tau_k)_{k \in \mathbb{N}}$ of predictable times that exhausts its jumps. Define next the bounded predictable process

$$\zeta = \mathbf{1}_{\{\nu^X(\{\cdot\}) > 0\}} = \sum_{k=1}^{\infty} \mathbf{1}_{[\tau_k]}.$$

Setting $X^{\text{qc}} = (1 - \zeta) \cdot X$ and $X^{\text{dp}} = \zeta \cdot X$ then yields the decomposition in (3.3), the quasi-left-continuity of X^{qc} , and $[X^{\text{qc}}, X^{\text{dp}}] = 0$. Finally, setting $D_n = (\Omega \times [0, \infty)) \setminus \bigcup_{k=n}^{\infty} [\tau_k]$ in Definition 3.1(i) for each $n \in \mathbb{N}$ yields $X^{\text{dp}} \in \mathcal{V}_\sigma^d$. \square

4. Classification of pure-jump processes

The following definition and theorem provide a precise formulation of the relationships among the various families of pure-jump processes. For notation and setup, see Section 2. For a review of the semimartingale topology on the space of semimartingales \mathcal{S} , see Appendix A.

Definition 4.1. *Consider the following subsets of \mathcal{S} .*

- \mathcal{J}^1 : the class of quadratic pure-jump processes, i.e., those semimartingales X that satisfy $[X, X]^c = [X, X] - x^2 \star \mu^X = 0$ (see Protter [11, p. 63]).
- \mathcal{J}^2 : the class of pure-jump processes, i.e., those semimartingales X that satisfy

$$X = X_0 + \sum_{k=1}^{\infty} \Delta X_{\tau_k} \mathbf{1}_{[\tau_k, \infty[} \quad (4.1)$$

in the semimartingale topology for a family $(\tau_k)_{k \in \mathbb{N}}$ of stopping times.

- \mathcal{J}^3 : the class of strong pure-jump processes, i.e., those semimartingales $X \in \mathcal{J}^2$ that satisfy $X = Y$ for all $Y \in \mathcal{J}^2$ with $\mu^Y = \mu^X$ and $Y_0 = X_0$.
- $\mathcal{J}^4 = \mathcal{V}_\sigma^d$: the sigma-localized class of finite variation pure-jump processes.
- $\mathcal{J}^5 = \mathcal{V}^d$: the class of finite variation pure-jump processes, i.e., those semimartingales X that satisfy $X = X_0 + x * \mu^X$.
- \mathcal{J}^6 : the class of piecewise constant processes with finitely many jumps on each finite time interval. \square

Remark 4.2. Let us explain (4.1) along with the qualifier “in the semimartingale topology.” To this end, for each $n \in \mathbb{N}$, consider the process $X^{(n)} = X_0 + \sum_{k=1}^n \Delta X_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}$. Then (4.1) should be read as $X = \lim_{n \uparrow \infty} X^{(n)}$ in the semimartingale topology, meaning that

$$\lim_{n \uparrow \infty} \left(\sup_{\zeta: |\zeta| \leq 1} \mathbb{E} \left[\left| \zeta \cdot X_t^{(k)} - \zeta \cdot X_t \right| \wedge 1 \right] \right) = 0$$

for all $t \geq 0$, where the supremum is taken over all predictable processes ζ with $|\zeta| \leq 1$. See, in particular, Definition A.1. One might consider other topologies than the semimartingale topology, for example the one induced by uniform convergence on compacts in probability. However, such a choice turns out to be impractical as illustrated in Appendix B. \square

Theorem 4.3. *We always have*

$$\mathcal{J}^1 \supsetneq \mathcal{J}^2 \supsetneq \mathcal{J}^3 \supsetneq \mathcal{J}^4 \supsetneq \mathcal{J}^5 \supsetneq \mathcal{J}^6. \quad (4.2)$$

In general, these set inclusions are strict. More precisely, there exists a filtered probability space such that simultaneously we have

$$\mathcal{J}^2 \supsetneq \mathcal{J}^3 \supsetneq \mathcal{J}^4 \supsetneq \mathcal{J}^5. \quad (4.3)$$

Moreover, the following statements hold.

- For all $i \in \{1, 2, 3, 4\}$, the families \mathcal{J}^i equal their sigma-localized class \mathcal{J}_σ^i ; that is, $\mathcal{J}^i = \mathcal{J}_\sigma^i$. Furthermore, by definition, $\mathcal{J}^4 = \mathcal{J}_\sigma^5$.
- For all $i \in \{1, 2, 3, 4, 6\}$, the families \mathcal{J}^i are closed under stochastic integration.
- For all $i \in \{1, \dots, 6\}$, the families \mathcal{J}^i are invariant under equivalent measure changes. More precisely, with the obvious notation, if \mathbf{Q} is locally absolutely continuous with respect to \mathbf{P} we have $\mathcal{J}^i(\mathbf{P}) \subset \mathcal{J}^i(\mathbf{Q})$ for all $i \in \{1, \dots, 6\}$.

Theorem 4.3 is proved in Section 5. The strictness of the inclusion $\mathcal{J}^4 \subsetneq \mathcal{J}^3$ is of interest. It says that there exist strong pure-jump processes, i.e., pure-jump processes uniquely determined by their jump measure, that are not sigma-locally of finite variation. To prove the strictness of this inclusion, Subsection 5.6 contains a specific example of such a process X (Example 5.17). This example relies on a jump measure μ^X with predictable compensator ν^X that supports a countable set of jump sizes.

To gain insight, consider the disintegrated form $\nu^X(dt, dx) = F_t^X(dx) dA_t^X$, where F^X is a transition kernel (see Section 2 for more details). The jump measure μ^X in Example 5.17 relies on a kernel F^X that has large atoms in a neighbourhood of zero. As it turns out, this example is canonical. Indeed, Corollary 4.4 below states if X is a strong pure-jump process

whose associated jump size kernel does not allow for too many large atoms, then X must be sigma-locally of finite variation.

We have already observed that a process $X \in \mathcal{J}^4 \subset \mathcal{J}^3$ is uniquely described by its jump measure. The following corollary of the proof of Theorem 4.3 provides explicit characterizations of the processes in \mathcal{J}^4 in relation to the bigger classes \mathcal{J}^1 , \mathcal{J}^2 , and \mathcal{J}^3 . A further analytic representation for such processes has been provided in Proposition 3.12.

Corollary 4.4. *Let X denote a process. Then the following statements are equivalent.*

- (i) $X \in \mathcal{J}^4 = \mathcal{V}_\sigma^d$.
- (ii) $X \in \mathcal{J}^3$ and

$$\left(\limsup_{x \downarrow 0} x F^X(\{x\}) \right) \wedge \left(\limsup_{x \uparrow 0} |x| F^X(\{x\}) \right) = 0, \quad (\mathbf{P} \times dA^X)\text{-a.e.} \quad (4.4)$$

- (iii) $X \in \mathcal{J}^2$ and

$$\int |x| \mathbf{1}_{|x| \leq 1} F^X(dx) < \infty, \quad (\mathbf{P} \times dA^X)\text{-a.e.} \quad (4.5)$$

- (iv) $X \in \mathcal{J}^1$, (4.5) holds, $\int_0^\cdot \int x \mathbf{1}_{|x| \leq 1} F_t^X(dx) dA_t^X < \infty$, and

$$B^{X[1]} = \int_0^\cdot \left(\int x \mathbf{1}_{|x| \leq 1} F_t^X(dx) \right) dA_t^X = x \mathbf{1}_{|x| \leq 1} \star \nu^X.$$

Here $B^{X[1]}$ denotes the drift of $X - x \mathbf{1}_{|x| > 1} \star \mu^X$; see also Section 2.

Corollary 4.4 is proved in Subsection 5.5. Note that the condition in (4.4) is satisfied, for example, if F^X is atomless $(\mathbf{P} \times dA^X)\text{-a.e.}$

The following is a corollary of the inclusion $\mathcal{J}^2 \subset \mathcal{J}^1$, stated in Theorem 4.3. If $X \in \mathcal{J}^1$ is predictable then it is well known that X is a finite-variation process. The next assertion illustrates that restricting oneself to \mathcal{J}^2 yields a finite-variation pure-jump process.

Corollary 4.5. *If $X \in \mathcal{J}^2$ is predictable then $X \in \mathcal{J}^5 = \mathcal{V}^d$.*

Proof. Since $X \in \mathcal{J}^2 \subset \mathcal{J}^1$ is predictable we have that X is a finite-variation process. But then the convergence in (4.1) is actually in finite-variation and X is the sum of its jumps, concluding the proof. \square

We conclude this section with a remark on the process $X^{\text{dp}} \in \mathcal{V}_\sigma^d$ of the semimartingale decomposition in Proposition 3.15. Observe that the family of predictable times $\mathcal{T} = (\tau_k)_{k \in \mathbb{N}}$ from the proof of Proposition 3.15 exhausts the jumps of X^{dp} . Simultaneously, Theorem 4.3 yields $X^{\text{dp}} \in \mathcal{J}^2$. A priori, it is not clear that \mathcal{T} is good enough to approximate X^{dp} in \mathcal{J}^2 because the membership of \mathcal{J}^2 only ever guarantees one exhausting sequence of stopping times (with the desired convergence property) and that sequence is not even predictable in principle. The next result therefore appears to be rather strong.

Proposition 4.6. *Let X satisfy $X = X^{\text{dp}}$ in the notation of Proposition 3.15. Let $(\tau_k)_{k \in \mathbb{N}}$ be any sequence of predictable times that exhausts the jumps of X . Then $(\tau_k)_{k \in \mathbb{N}}$ also approximates X in \mathcal{J}^2 , i.e., we have $X = \sum_{k=1}^\infty \Delta X_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}$ in the semimartingale topology.*

Proof. Apply Lemma A.3(iii) below with the same sequence $(D_n)_{n \in \mathbb{N}}$ as in the proof of Proposition 3.15. \square

For a related statement about the summability of jumps of a semimartingale $X = X^{\text{dp}}$ at predictable times, see Galtchouk [5]. There it is shown that for any sequence $(\tau_k)_{k \in \mathbb{N}}$ of predictable times that exhausts the jumps of X the limit $\lim_{n \uparrow \infty} \sum_{k=1}^n \Delta X_{\tau_k} \mathbf{1}_{\{\tau_k \leq t\}}$ exists almost surely for each $t \geq 0$. (Only the case when X is a local martingale is considered, however the extension to the general case is straightforward.)

5. Proof of Theorem 4.3 (and of Corollary 4.4)

This section contains the proof of this paper's main theorem. It is split up in six subsections. Subsections 5.1, 5.2, and 5.3 provide the proofs of Theorem 4.3(i), (ii), and (iii), respectively. Subsection 5.4 yields the set inclusions in (4.2). Then Subsection 5.5 focuses on the proof of Corollary 4.4, while Subsection 5.6 concludes with a proof of (4.3), namely the strictness of the inclusions.

5.1. Proof of Theorem 4.3(i)

In this subsection we argue that $\mathcal{J}^i = \mathcal{J}_\sigma^i$ for all $i \in \{1, 2, 3, 4\}$. Indeed, fix $X \in \mathcal{J}_\sigma^1$ and the corresponding sigma-localizing sequence $(D_k)_{k \in \mathbb{N}}$ of predictable sets. Then

$$[X, X]^c = \left(\lim_{k \uparrow \infty} \mathbf{1}_{D_k} \right) \cdot [X, X]^c = \lim_{k \uparrow \infty} (\mathbf{1}_{D_k} \cdot [X, X]^c) = \lim_{k \uparrow \infty} [\mathbf{1}_{D_k} \cdot X, \mathbf{1}_{D_k} \cdot X]^c = 0,$$

which yields $X \in \mathcal{J}^1$. As $\mathcal{J}^5 = \mathcal{V}^{\text{d}}$ is stable under sigma-stopping, [7, Proposition 2.1] yields the statement for $i = 4$.

The cases $i = 2$ and $i = 3$ follow from Lemmata 5.3 and 5.4. Before stating and proving them, we first present a useful tool for pure-jump processes in the next lemma.

Lemma 5.1. *Let $(X^{(k)})_{k \in \mathbb{N}} \subset \mathcal{J}^2$ be a sequence of pure-jump processes such that $X_0^{(k)} = 0$ and $[X^{(k)}, X^{(l)}] = 0$ for all $k, l \in \mathbb{N}$ with $k \neq l$. Then the following two statements are equivalent.*

- (I) $\sum_{k=1}^{\infty} [X^{(k)}, X^{(k)}] < \infty$ and $\sum_{k=1}^{\infty} B^{X^{(k)}}[1]$ converges in the \mathcal{S} -topology to a process B .
- (II) $\sum_{k=1}^{\infty} X^{(k)}$ converges in the \mathcal{S} -topology to a process X .

If one (hence both) of these conditions hold then $B^{X[1]} = B$, $\sum_{k=1}^{\infty} \Delta X^{(k)} = \Delta X$, and X is a pure-jump process.

Proof. Thanks to Lemma A.4(iii) it suffices to argue that X is a pure-jump process provided the two statements hold. For each $k \in \mathbb{N}$ we have a sequence $(\tau_n^{(k)})_{n \in \mathbb{N}}$ of stopping times (by possibly setting $\tau_n^{(k)} = \infty$ for n large enough if $X^{(k)}$ has only finitely many jumps) such that $X^{(k)} = \sum_{n=1}^{\infty} \Delta X_{\tau_n^{(k)}}^{(k)} \mathbf{1}_{\llbracket \tau_n^{(k)}, \infty \rrbracket}$ in the \mathcal{S} -topology and $\Delta X_{\tau_n^{(k)}}^{(k)} \neq 0$ on $\{\tau_n^{(k)} < \infty\}$. Thanks to Lemma A.4(iii) we have $\Delta X_{\tau_n^{(k)}} = \Delta X_{\tau_n^{(k)}}^{(k)}$ on $\{\tau_n^{(k)} < \infty\}$ for all $k, n \in \mathbb{N}$. Furthermore, $(\tau_n^{(k)})_{k, n \in \mathbb{N}}$ exhausts the jumps of X .

Next, for each $m \in \mathbb{N}$, let K_m and N_m be the smallest integers such that

$$d_{\mathcal{S}} \left(X, \sum_{k=1}^{K_m} X^{(k)} \right) \leq \frac{1}{2m}$$

and

$$d_{\mathcal{S}} \left(X^{(k)}, \sum_{n=1}^{N_m} \Delta X_{\tau_n^{(k)}}^{(k)} \mathbf{1}_{[\tau_n^{(k)}, \infty[} \right) \leq \frac{1}{2mK_m} \quad \text{for all } k \in \{1, \dots, K_m\}.$$

By a standard diagonalization argument we can now construct a sequence of stopping times $(\tau_i)_{i \in \mathbb{N}}$ such that

$$\lim_{m \uparrow \infty} d_{\mathcal{S}} \left(X, \sum_{i=1}^m \Delta X_{\tau_i} \mathbf{1}_{[\tau_i, \infty[} \right) = 0,$$

yielding the statement. \square

Corollary 5.2. *The sum of two pure-jump processes whose quadratic covariation is zero is again a pure-jump process.*

The next two lemmata exploit the fact that \mathcal{J}^2 is stable under sigma-stopping thanks to Lemma A.3(ii).

Lemma 5.3. *If $X \in \mathcal{S}$ is sigma-locally a pure-jump process then it is a pure-jump process.*

Proof. By assumption there exists a nondecreasing sequence $(D_k)_{k \in \mathbb{N}}$ of predictable sets such that $\bigcup_{k \in \mathbb{N}} D_k = \Omega \times [0, \infty)$ and $\mathbf{1}_{D_k} \cdot X$ is a pure-jump process. With $D_0 = \emptyset$, define $\overline{D}_k = D_k \setminus D_{k-1}$ for all $k \in \mathbb{N}$ and note that $X^{(k)} = \mathbf{1}_{\overline{D}_k} \cdot X$ is also a pure-jump process as \mathcal{J}^2 is stable under sigma-stopping. Moreover, we have $[X^{(k)}, X^{(l)}] = 0$ for all $k, l \in \mathbb{N}$ with $k \neq l$ and

$$\sum_{k=1}^n X^{(k)} = \sum_{k=1}^n (\mathbf{1}_{\overline{D}_k} \cdot X) = \mathbf{1}_{D_n} \cdot X, \quad n \in \mathbb{N},$$

which converges in the \mathcal{S} -topology to X (as $n \uparrow \infty$) thanks to Lemma A.3(iii). Hence by Lemma 5.1, X is a pure-jump process. \square

Lemma 5.4. *If $X \in \mathcal{S}$ is sigma-locally a strong pure-jump process then it is a strong pure-jump process.*

Proof. By assumption there exists a nondecreasing sequence $(D_k)_{k \in \mathbb{N}}$ of predictable sets such that $\bigcup_{k \in \mathbb{N}} D_k = \Omega \times [0, \infty)$ and $\mathbf{1}_{D_k} \cdot X$ is a strong pure-jump process. Assume there exists a pure-jump process Y with $\mu^Y = \mu^X$ and $Y_0 = X_0$ but $Y \neq X$. Since $\lim_{k \uparrow \infty} (\mathbf{1}_{D_k} \cdot X) = X - X_0$ and $\lim_{k \uparrow \infty} (\mathbf{1}_{D_k} \cdot Y) = Y - Y_0$ in the \mathcal{S} -topology thanks to Lemma A.3(iii), there exists some $k \in \mathbb{N}$ such that $\mathbf{1}_{D_k} \cdot X \neq \mathbf{1}_{D_k} \cdot Y$. This, however, contradicts the assumption since $\mathbf{1}_{D_k} \cdot Y \in \mathcal{J}^2$ for each $k \in \mathbb{N}$ because \mathcal{J}^2 is stable under sigma-stopping. \square

5.2. Proof of Theorem 4.3(ii)

In this subsection we argue that \mathcal{J}^i is closed under stochastic integration for all $i \in \{1, 2, 3, 4, 6\}$. First, the cases $i = 1$ and $i = 6$ are clear. The case $i = 4$ follows from Corollary 3.14.

For the case $i = 2$, assume that $X \in \mathcal{J}^2$ and fix $\zeta \in L(X)$. We need to argue that $\zeta \cdot X \in \mathcal{J}^2$. Thanks to Theorem 4.3(i) (see also Lemma 5.3), we may assume that $|\zeta|$ is bounded. The statement then follows directly from the definition of \mathcal{S} -topology.

The remaining case $i = 3$ follows from the next lemma.

Lemma 5.5. *Let $X \in \mathcal{J}^3$ and $\zeta \in L(X)$. If $Y \in \mathcal{J}^2$ is a pure-jump process with $\mu^Y = \mu^{\zeta \cdot X}$ then $Y = \zeta \cdot X$.*

Proof. Note that

$$Z = \left(\mathbf{1}_{\{\zeta \neq 0\}} \frac{1}{\zeta} \right) \cdot Y + \mathbf{1}_{\{\zeta = 0\}} \cdot X$$

satisfies $\mu^Z = \mu^X$. Moreover, Z is a pure-jump process thanks to Corollary 5.2 in conjunction with the closedness of \mathcal{J}^2 under stochastic integration. Since $X \in \mathcal{J}^3$ we get $Z = X$. This again yields that

$$Y = \mathbf{1}_{\{\zeta = 0\}} \cdot Y + \mathbf{1}_{\{\zeta \neq 0\}} \cdot Y = \mathbf{1}_{\{\zeta = 0\}} \cdot Y + \zeta \cdot Z = \mathbf{1}_{\{\zeta = 0\}} \cdot Y + \zeta \cdot X.$$

We conclude after observing that $\mu^{\mathbf{1}_{\{\zeta = 0\}} \cdot Y} = 0$ and $\mathbf{1}_{\{\zeta = 0\}} \cdot Y \in \mathcal{J}^2$, hence $\mathbf{1}_{\{\zeta = 0\}} \cdot Y = 0$. The last step relies on the fact that if $X \in \mathcal{J}^2$ and $\mu^X = 0$ then $X = 0$; i.e., a pure-jump process that has no jumps has to equal the zero process. \square

5.3. Proof of Theorem 4.3(iii)

In this subsection we argue that if \mathbf{Q} is locally absolutely continuous with respect to \mathbf{P} we have $\mathcal{J}^i(\mathbf{P}) \subset \mathcal{J}^i(\mathbf{Q})$ for all $i \in \{1, \dots, 6\}$. The cases $i = 1$, $i = 5$, and $i = 6$ are clear. The case $i = 2$ follows from Lemma A.3(vi). The case $i = 4$ is a consequence of Proposition 3.12 and Remark 3.2.

The remaining case $i = 3$ follows from Lemma 5.7. It requires the following result regarding the lift of a pure-jump process from $\mathcal{J}^2(\mathbf{Q})$ to $\mathcal{J}^2(\mathbf{P})$.

Lemma 5.6. *Let Z denote a nonnegative \mathbf{P} -martingale, \mathbf{Q} a probability measure that satisfies $d\mathbf{Q}/d\mathbf{P}|_{\mathcal{F}_t} = Z_t$ for all $t \geq 0$, and Y an element of $\mathcal{J}^2(\mathbf{Q})$. Assume that there exists some stopping time σ such that $Y = Y^\sigma$, \mathbf{P} -almost surely, and Z^σ does not hit zero continuously, \mathbf{P} -almost surely. Then there exists a \mathbf{P} -semimartingale $Y_\uparrow \in \mathcal{J}^2(\mathbf{P})$ with $Y_\uparrow = Y$, \mathbf{Q} -almost surely, and $Y_\uparrow = Y_\uparrow^\sigma$, \mathbf{P} -almost surely.*

Proof. Let $(\tau_k)_{k \in \mathbb{N}}$ denote a sequence of stopping times such that $Y = Y_0 + \sum_{k=1}^{\infty} \Delta Y_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}$ in the \mathcal{S} -topology under \mathbf{Q} . We may now assume that $Y_0 = 0$ and $\{\tau_k < \infty\} \cap \{Z_{\tau_k} = 0\} = \emptyset$ for all $k \in \mathbb{N}$. With these assumptions in place, we have $[\Delta Y_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}, \Delta Y_{\tau_l} \mathbf{1}_{[\tau_l, \infty[}] = 0$ for all $k, l \in \mathbb{N}$ with $k \neq l$ and $\sum_{k=1}^{\infty} [\Delta Y_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}, \Delta Y_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}] < \infty$, \mathbf{P} -almost surely. Consider next the sum $B^{(n)} = \sum_{k=1}^n B^{\Delta Y_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}]}(\mathbf{P})$ for each $n \in \mathbb{N}$. Suppose for the moment that $(B^{(n)})_{n \in \mathbb{N}}$ converges in the \mathcal{S} -topology under \mathbf{P} . Then Lemma 5.1 yields a limiting process

$Y_\uparrow \in \mathcal{J}^2(\mathbb{P})$ and Lemma A.3(vi) yields that $Y_\uparrow = Y$, \mathbb{Q} -almost surely. Clearly, we then also have $Y_\uparrow = Y_\uparrow^\sigma$, \mathbb{P} -almost surely.

We still need to argue the convergence of the \mathbb{P} -drifts $(B^{(n)})_{n \in \mathbb{N}}$ in the \mathcal{J} -topology under \mathbb{P} . To this end, we will first argue that $\lim_{n \uparrow \infty} B^{(n)} = B$ in the \mathcal{J} -topology under \mathbb{Q} for some predictable, \mathbb{Q} -almost surely finite variation process B . To see this, note that $Y \in \mathcal{J}^2(\mathbb{Q})$ and Lemma A.4(i) in conjunction with Lemma A.3(vi) yield that

$$\left(\sum_{k=1}^n \Delta Y_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}[1] \right)_{n \in \mathbb{N}} \quad \text{and} \quad \left(\sum_{k=1}^n \Delta Y_{\tau_k} \mathbf{1}_{[\tau_k, \infty[}[1] - B^{(n)} \right)_{n \in \mathbb{N}}$$

both converge in the \mathcal{J} -topology under \mathbb{Q} . This yields the convergence of $(B^{(n)})_{n \in \mathbb{N}}$ in the \mathcal{J} -topology under \mathbb{Q} to some process B , which may be chosen predictable by Lemma A.3(v). By Lemma A.3(iv), we have $[B, B]^c = 0$, hence B is of finite variation under \mathbb{Q} . We also also have $B = B^\sigma$, \mathbb{Q} -almost surely, and may assume that $B = B^\sigma$, \mathbb{P} -almost surely. Similarly, we may assume that B has right-continuous paths, \mathbb{P} -almost surely.

Note that the first time σ' that B is of infinite variation is predictable since the total-variation process of B is predictable; hence $\mathbb{E}_{\sigma', -}^\mathbb{P}[\Delta Z_{\sigma'}] = 0$ on $\{\sigma' < \infty\}$. Since B is of finite variation, \mathbb{Q} -almost surely, we conclude that $\sigma' = \infty$, \mathbb{P} -almost surely. Hence B is of finite variation, \mathbb{P} -almost surely, and has left limits. It suffices to argue now that the variations of $(B - B^{(n)})_{n \in \mathbb{N}}$ converge to zero in probability under \mathbb{P} . As they do under \mathbb{Q} and as the first time σ'' that the variations of $(B - B^{(n)})_{n \in \mathbb{N}}$ do not converge to zero is predictable, we may argue again that $\sigma'' = \infty$, \mathbb{P} -almost surely. This yields $\lim_{n \uparrow \infty} B^{(n)} = B$ in the \mathcal{J} -topology under \mathbb{P} , concluding the proof. \square

Lemma 5.7. *Let \mathbb{Q} be a probability measure that is locally absolutely continuous with respect to \mathbb{P} and let $X \in \mathcal{J}^3(\mathbb{P})$. If $Y \in \mathcal{J}^2(\mathbb{Q})$ is a pure-jump process with $\mu^Y = \mu^X$, \mathbb{Q} -almost surely, then $Y = X$, \mathbb{Q} -almost surely.*

Proof. For each $m \in \mathbb{N}$, let σ_m be the first time that the nonnegative martingale $(d\mathbb{Q}/d\mathbb{P}|_{\mathcal{F}_t})_{t \geq 0}$ crosses the level $1/m$. Then $X^{\sigma_m} \in \mathcal{J}^3(\mathbb{P})$ for each $m \in \mathbb{N}$ by Lemma 5.5 and $\lim_{m \uparrow \infty} \sigma_m = \infty$, \mathbb{Q} -almost surely. Hence, thanks to Lemma 5.4 applied under \mathbb{Q} , we may and shall assume from now on that $X = X^\sigma$, where $\sigma = \sigma_m$ for some $m \in \mathbb{N}$. Let $Y \in \mathcal{J}^2(\mathbb{Q})$ satisfy $\mu^Y = \mu^X$, \mathbb{Q} -almost surely, and $Y_0 = X_0$. Then $Y = Y^\sigma$ and Lemma 5.6 yields $Y_\uparrow \in \mathcal{J}^2(\mathbb{P})$ with $Y_\uparrow = Y$, \mathbb{Q} -almost surely, and $Y_\uparrow = Y_\uparrow^\sigma$, \mathbb{P} -almost surely.

Define now the \mathbb{P} -semimartingale $Y' = Y_\uparrow + (\Delta X_\sigma - \Delta Y_{\uparrow\sigma}) \mathbf{1}_{[\sigma, \infty[}$. Then we have $\mu^{Y'} = \mu^X$ and $Y' = Y$, \mathbb{Q} -almost surely. Since $X \in \mathcal{J}^3(\mathbb{P})$ we thus have $Y' = X$, \mathbb{P} -almost surely, yielding $Y = Y' = X$, \mathbb{Q} -almost surely, as required. \square

5.4. Proof of the set inclusions in (4.2)

Lemma A.3(iv) yields the inclusion $\mathcal{J}^1 \supset \mathcal{J}^2$. The strictness of this inclusion follows from observing, for example, that $X_t = t$ for all $t \geq 0$ satisfies $X \in \mathcal{J}^1 \setminus \mathcal{J}^2$. The inclusions $\mathcal{J}^2 \supset \mathcal{J}^3$ and $\mathcal{J}^4 \supset \mathcal{J}^5 \supset \mathcal{J}^6$ are clear. Since the deterministic semimartingale $X = \sum_{k=1}^\infty k^{-2} \mathbf{1}_{[1/k, \infty[}$ satisfies $X \in \mathcal{J}^5 \setminus \mathcal{J}^6$, we also have the strictness of the last inclusion.

To see $\mathcal{J}^3 \supset \mathcal{J}^4$, consider now $X \in \mathcal{J}^5$. By definition of the \mathcal{J} -topology every exhausting sequence for X also yields an approximating sequence of stopping times for X in \mathcal{J}^2 . This shows $\mathcal{J}^5 \subset \mathcal{J}^2$, and in fact $\mathcal{J}^5 \subset \mathcal{J}^3$. Hence Lemma 5.4 yields $\mathcal{J}^4 = \mathcal{J}_\sigma^5 \subset \mathcal{J}_\sigma^3 = \mathcal{J}^3$.

5.5. Proof of Corollary 4.4 and the strictness of the inclusion

$$\mathcal{J}^2 \supset \mathcal{J}^3$$

Let us outline here the proof.

- Assume that Corollary 4.4(i) holds. Then Proposition 3.12 yields that $x \in L_\sigma(\mu^X)$. Thanks to Proposition 3.4 we moreover have $x\mathbf{1}_{|x|<1} \in L_\sigma(\nu^X)$. Lemma 3.3, the set inclusions of (4.2), proven in Subsection 5.4, and the representation in (3.2) yield that Corollary 4.4(ii)–(iv) also hold.
- Assume that Corollary 4.4(iv) holds. Thanks to Proposition 3.4 and Lemma 3.3 we then have $x \in L_\sigma(\mu^X)$ and

$$x \star \mu^X = x\mathbf{1}_{|x|>1} \star \mu^X + x\mathbf{1}_{|x|\leq 1} \star (\mu^X - \nu^X) + x\mathbf{1}_{|x|\leq 1} \star \nu^X.$$

However, the right hand side of the last display is exactly $X - X_0$, thanks to the representation in [6, II.2.34] for quadratic pure-jump processes X , namely

$$X = X_0 + x\mathbf{1}_{|x|>1} \star \mu^X + x\mathbf{1}_{|x|\leq 1} \star (\mu^X - \nu^X) + B^{X[1]}. \quad (5.1)$$

Hence $X - X_0 = x \star \mu^X$ and Proposition 3.12 then yields the implication from (iv) to (i) in Corollary 4.4.

- Assume that Corollary 4.4(iii) holds. Then Lemma 5.8(i) below yields that Corollary 4.4(i) also holds.
- Assume that Corollary 4.4(ii) holds. Then Lemma 5.11 below yields that Corollary 4.4(i) also holds.
- Finally, Lemma 5.12 shows that the inclusion $\mathcal{J}^2 \supset \mathcal{J}^3$ is usually strict.

Lemma 5.8. *Let $X \in \mathcal{J}^2$ be a pure-jump semimartingale and define the predictable set*

$$D = \left\{ (\omega, t) : \int |x| \mathbf{1}_{|x|\leq 1} F_t^X(dx) < \infty \right\}.$$

Then the following statements hold.

- (i) *We have $\mathbf{1}_D \cdot X \in \mathcal{V}_\sigma^d$.*
- (ii) *The following equalities hold.*

$$\begin{aligned} D &= \left\{ (\omega, t) : \int x^+ \mathbf{1}_{x^+ \leq 1} F_t^X(dx) < \infty \right\} \\ &= \left\{ (\omega, t) : \int x^- \mathbf{1}_{x^- \leq 1} F_t^X(dx) < \infty \right\}, \quad (\mathbf{P} \times dA^X)\text{-a.e.} \end{aligned}$$

- (iii) *There exists a predictable process β^X with $|\beta^X| \cdot A^X < \infty$ such that $B^{X[1]} = \beta^X \cdot A^X$. Moreover, on D we have $\beta^X = \int x \mathbf{1}_{|x|\leq 1} F_t^X(dx)$, $(\mathbf{P} \times dA^X)\text{-a.e.}$*

Proof. For (i), thanks to Theorem 4.3(ii) (or the fact that \mathcal{J}^2 is stable under sigma-stopping), we may assume that $X = \mathbf{1}_D \cdot X$. Define now the predictable sets

$$D_k = \left\{ (\omega, t) : \int |x| \mathbf{1}_{|x|\leq 1} F_t^X(dx) \leq k \right\}, \quad k \in \mathbb{N},$$

and $\overline{D}_k = D_k \setminus D_{k-1}$ with $D_0 = \emptyset$. Again $X^{(k)} = \mathbf{1}_{\overline{D}_k} \cdot X$ is a pure-jump process for each $k \in \mathbb{N}$. Moreover, by Proposition 3.4 and Lemma 3.3, $x \in L_\sigma(\mu^{X^{(k)}})$ for each $k \in \mathbb{N}$. Since by (4.2) $x \star \mu^{X^{(k)}} \in \mathcal{J}^4 \subset \mathcal{J}^3$ we have $X^{(k)} = x \star \mu^{X^{(k)}}$ for each $k \in \mathbb{N}$. Thanks to Proposition 3.12 this yields

$$B^{X^{(k)}[1]} = \int_0^\cdot \mathbf{1}_{\overline{D}_k}(t) \left(\int x \mathbf{1}_{|x| \leq 1} F_t^X(dx) \right) dA_t^X, \quad k \in \mathbb{N}.$$

Hence by Lemmata A.3(iii) and 5.1 we have

$$\int_0^\cdot \mathbf{1}_{D_n}(t) \left(\int x \mathbf{1}_{|x| \leq 1} F_t^X(dx) \right) dA_t^X$$

converges in the \mathcal{S} -topology (as $n \uparrow \infty$) to a finite variation process, yielding the result.

To see (ii), fix $\kappa > 0$ and consider the predictable set

$$D' = \left\{ (\omega, t) : \int x^- \mathbf{1}_{x^- \leq 1} F_t^X(dx) < \kappa \right\}.$$

By symmetry, it suffices now to argue that $\mathbf{1}_{D'} x^+ \mathbf{1}_{x^+ \leq 1} \nu^X < \infty$. As above, we may assume that $X = \mathbf{1}_{D'} \cdot X$. Then $x^- \star \mu^X < \infty$ and (A.1) with $\zeta = 1$ yield that also $x^+ \star \mu^X < \infty$; hence $x^+ \mathbf{1}_{x^+ \leq 1} \star \nu^X < \infty$ as required.

For (iii) Lebesgue's decomposition of measures yields a predictable set \tilde{D} and a predictable process β^X such that $\mathbf{1}_{\tilde{D}} \cdot A^X = 0$, $|\beta^X| \cdot A^X < \infty$, and $B^{X[1]} = \mathbf{1}_{\tilde{D}} \cdot B^{X[1]} + \beta^X \cdot A^X$. Next, note that $\mathbf{1}_{\tilde{D}} \cdot X$ is also a pure-jump process and $x^2 \star \nu^{\mathbf{1}_{\tilde{D}} \cdot X[1]} = \mathbf{1}_{\tilde{D}} x^2 \star \nu^{X[1]} = 0$, yielding $\mathbf{1}_{\tilde{D}} \cdot X[1] = 0$, hence $\mathbf{1}_{\tilde{D}} \cdot B^{X[1]} = 0$. The second assertion of (iii) follows directly from (i) and Proposition 3.12. \square

Lemma 5.9. *Let μ be a jump measure with $x^2 \star \mu < \infty$. Moreover, let $(f^{(k)})_{k \in \mathbb{N}}$ and $(g^{(k)})_{k \in \mathbb{N}}$ denote two nonincreasing sequences of strictly positive predictable processes. For each $k \in \mathbb{N}$, define*

$$\beta^{(k)} = \int x \mathbf{1}_{\{x \in [-1, -g^{(k)}] \cup [f^{(k)}, 1]\}} F(dx).$$

If $|\beta^{(k)}| \cdot A < \infty$ for each $k \in \mathbb{N}$ and $\beta^{(k)} \cdot A$ converges in the \mathcal{S} -topology (as $k \uparrow \infty$) to a process B , then there exists a pure-jump process X with $\mu^X = \mu$ and $B^{X[1]} = B$.

Proof. Note that

$$\int |x| \mathbf{1}_{\{x \in [-1, -g^{(k)}] \cup [f^{(k)}, 1]\}} F(dx) < \infty, \quad (\mathbf{P} \times dA)\text{-a.e.}, \quad k \in \mathbb{N},$$

by the strict positivity of $g^{(k)}$ and $f^{(k)}$. Hence by assumption and Propositions 3.4 and 3.12, $x \mathbf{1}_{|x|=1}, x \mathbf{1}_{\{x \in (-g^{(k-1)}, -g^{(k)}] \cup [f^{(k)}, f^{(k-1)})\}} \in L_\sigma(\mu)$, $X^{(0)} = x \mathbf{1}_{|x|=1} \star \mu \in \mathcal{V}_\sigma^d \subset \mathcal{J}^2$, and

$$X^{(k)} = x \mathbf{1}_{\{x \in (-g^{(k-1)}, -g^{(k)}] \cup [f^{(k)}, f^{(k-1)})\}} \star \mu \in \mathcal{V}_\sigma^d \subset \mathcal{J}^2, \quad k \in \mathbb{N},$$

where $g^{(0)} = f^{(0)} = 1$. An application of Lemma 5.1 now concludes. \square

The following lemma complements Lemma 5.8(ii)&(iii). Given a jump measure μ and a predictable process β , both satisfying technical conditions, it constructs a pure-jump process X with $\mu^X = \mu$ and drift rate β on the predictable set where the jump sizes do not integrate.

Lemma 5.10. *Let μ be a jump measure with $x^2 * \mu < \infty$ and*

$$\left(\limsup_{x \downarrow 0} xF(\{x\}) \right) \wedge \left(\limsup_{x \uparrow 0} |x|F(\{x\}) \right) = 0, \quad (\mathbf{P} \times dA)\text{-a.e.} \quad (5.2)$$

Assume that the predictable set

$$D = \left\{ (\omega, t) : \int |x| \mathbf{1}_{|x| \leq 1} F_t(dx) < \infty \right\}$$

satisfies

$$\begin{aligned} D &= \left\{ (\omega, t) : \int x^+ \mathbf{1}_{x^+ \leq 1} F_t(dx) < \infty \right\} \\ &= \left\{ (\omega, t) : \int x^- \mathbf{1}_{x^- \leq 1} F_t(dx) < \infty \right\}, \quad (\mathbf{P} \times dA)\text{-a.e.} \end{aligned} \quad (5.3)$$

and let β denote any nonnegative predictable process such that

$$\int_0^\cdot \mathbf{1}_{D^c}(t) |\beta_t| dA_t < \infty.$$

Then there exists a pure-jump process $X \in \mathcal{J}^2$ such that $\mu^X = \mu$ and

$$B^{X[1]} = \int_0^\cdot \left(\mathbf{1}_D(t) \int x \mathbf{1}_{|x| \leq 1} F_t(dx) + \mathbf{1}_{D^c}(t) \beta_t \right) dA_t.$$

Proof. Consider the predictable sets

$$D' = D^c \cap \left\{ (\omega, t) : \limsup_{x \downarrow 0} xF_t(\{x\}) = 0 \right\}; \quad D'' = D^c \setminus D'.$$

By Corollary 5.2, symmetry, and Subsection 3.3 we may assume that $D'' = \emptyset$, $\mathbf{1}_{|x| > 1} * \mu = 0$, $\mu = \mathbf{1}_{D'} \mu$, and $\beta = \mathbf{1}_{D'} \beta$.

To make headway, consider the predictable process

$$c = \inf \{ \varepsilon > 0 : \varepsilon F(\{\varepsilon\}) > 1 \} \wedge 2$$

and note that $c > 0$ by assumption. Next, consider the process

$$d = \mathbf{1}_{D'^c} + \mathbf{1}_{D'} \sup \left\{ \varepsilon > 0 : \beta + \int |x| \mathbf{1}_{x \in [-1, -\varepsilon]} F(dx) \geq \int x \mathbf{1}_{\{x \in [c, 1]\}} F(dx) \right\}.$$

Then by (5.3), $d > 0$ and since

$$\{(\omega, t) : d_t \geq \varepsilon\} = \left\{ (\omega, t) : \int |x| \mathbf{1}_{x \in [-1, -\varepsilon]} F_t(dx) \geq \int x \mathbf{1}_{\{x \in [c, 1]\}} F_t(dx) - \beta \right\} \in \mathcal{P},$$

for all $\varepsilon \in (0, 1]$, it is easy to see that d is predictable. Next, define the processes

$$g^{(k)} = d \wedge \frac{1}{k}, \quad k \in \mathbb{N};$$

$$f^{(k)} = \mathbf{1}_{D'^c} + \mathbf{1}_{D'} \sup \left\{ \varepsilon > 0 : \int x \mathbf{1}_{x \in [\varepsilon, 1]} F(dx) \geq \beta + \int |x| \mathbf{1}_{\{x \in [-1, -g^{(k)}]\}} F(dx) \right\}, \quad k \in \mathbb{N}.$$

Similarly as for d we may argue that $f^{(k)} > 0$ and $f^{(k)}$ is predictable for each $k \in \mathbb{N}$. Again by (5.3), we also have $\lim_{k \uparrow \infty} f^{(k)} = 0$ on D' . Note that we also have $f^{(k)} < c$ for each $k \in \mathbb{N}$.

Next, define

$$\beta^{(k)} = \int x \mathbf{1}_{\{x \in [-1, -g^{(k)}] \cup [f^{(k)}, 1]\}} F(dx), \quad k \in \mathbb{N}.$$

Since

$$\beta^{(k)} \in \left[\beta, \beta + f^{(k)} F\left(\left\{f^{(k)}\right\}\right) \right] \subset [\beta, \beta + 1], \quad k \in \mathbb{N},$$

we have $|\beta^{(k)}| \cdot A < \infty$ and

$$\int_0^\cdot \beta^{(k)} dA_t \in \left[\int_0^\cdot \beta_t dA_t, \int_0^\cdot \left(\beta_t + f_t^{(k)} F_t\left(\left\{f_t^{(k)}\right\}\right) \right) dA_t \right], \quad k \in \mathbb{N}.$$

Since $\lim_{k \uparrow \infty} f^{(k)} = 0$ on D' we also have

$$\lim_{k \uparrow \infty} f^{(k)} F\left(\left\{f^{(k)}\right\}\right) = 0$$

by assumption. Hence, dominated convergence yields $\lim_{k \uparrow \infty} \beta^{(k)} \cdot A = \beta \cdot A$ in the \mathcal{S} -topology. An application of Lemma 5.9 now concludes. \square

Lemma 5.11. *Let $X \in \mathcal{J}^3$ denote a strong pure-jump process such that (4.4) holds. Then $X \in \mathcal{J}^4$.*

Proof. We only need to argue that (4.5) holds. Recall Lemma 5.8(ii) and apply Lemma 5.10 with $\mu = \mu^X$ and $\tilde{\beta} = 1$ and $\bar{\beta} = -1$. If (4.5) did not hold then we would obtain two pure-jump processes \tilde{X} and \bar{X} with $\tilde{X} \neq \bar{X}$ but with the same jump measures, contradicting the fact that $X \in \mathcal{J}^3$. This concludes the proof. \square

Lemma 5.12. *Assume that the filtered probability space is large enough so that it supports a probability measure μ that satisfies (5.2), (5.3), and*

$$\mathbb{P} \left[\int_0^\cdot \mathbf{1}_{\{\int |x| \mathbf{1}_{|x| \leq 1} F_t(dx) = \infty\}} dA_t > 0 \right] > 0.$$

Then $\mathcal{J}^2 \neq \mathcal{J}^3$.

Proof. As in the proof of Lemma 5.11, consider the two predictable processes $\tilde{\beta} = 1$ and $\bar{\beta} = -1$ and conclude by applying Lemma 5.10 twice. \square

As an illustration of Lemma 5.12 and a preparation for the next subsection, let us now discuss the Lévy situation by means of the following example. When X is a Lévy process we abuse the notation to treat F^X as a deterministic measure over \mathbb{R} rather than a stochastic process.

Example 5.13. Let X be an α -stable Lévy process without Brownian component. Specifically, take $F^X(dx) = \mathbf{1}_{x \neq 0}|x|^{-1-\alpha}dx$ for all $x \in \mathbb{R}$ with $0 < \alpha < 2$ and $A_t^X = t$ for all $t \geq 0$. Observe that $X \in \mathcal{J}^4$ is equivalent to $X \in \mathcal{J}^5$ since X is Lévy. Let us now write $\beta = B_t^{X[1]}/t$, where $t > 0$, for the drift rate of X .

- If $0 < \alpha < 1$ and $\beta = \int x \mathbf{1}_{|x| \leq 1} F^X(dx) = 0$ then X belongs to $\mathcal{J}^5 \setminus \mathcal{J}^6$.
- If $0 < \alpha < 1$ and $\beta \neq \int x \mathbf{1}_{|x| \leq 1} F^X(dx)$ then X belongs to $\mathcal{J}^1 \setminus \mathcal{J}^2$.
- If $1 \leq \alpha < 2$ then X belongs to $\mathcal{J}^2 \setminus \mathcal{J}^3$ for any value of β . □

5.6. Proof of (4.3)

On finite probability spaces we have $\mathcal{J}^2 = \mathcal{J}^5$. However, in general, this is not true. Lemma 5.12 already asserts that $\mathcal{J}^2 \neq \mathcal{J}^3$ as long as the probability space is large enough. The process X of the introduction shows that usually we have $\mathcal{J}^4 \neq \mathcal{J}^5$. Example 5.17 below illustrates that $\mathcal{J}^3 \neq \mathcal{J}^4$ is also possible.

Theorem 4.3 asserts that all these inequalities may hold simultaneously for some probability space. To see that such a probability space exists it suffices to piece together these three examples. For example, take the product of a probability space that allows for a process as in Example 5.17 and another probability space that satisfies the assumptions of Lemma 5.12 and additionally allows for a process as in the introduction. As filtration consider the one of Example 5.17 between time 0 and 1 and afterwards allow the filtration to be large enough to allow for the other examples.

Example 5.17 requires a few technical prerequisites that we introduce now. Throughout this subsection, \mathfrak{F}^X and \mathfrak{F}_+^X shall denote the natural filtration of a process X and its right-continuous modification. Additionally, for purely technical reasons, throughout this subsection we also assume to be on the canonical space as in Dellacherie and Meyer [2, Definition IV.95].

Lemma 5.14. Let g denote a $\{0, 1\}$ -valued \mathfrak{F}^X -predictable function and $X \in \mathcal{S}$ a semi-martingale. Then $g(\Delta X)$ is \mathfrak{F}^X -optional. Moreover, if τ is an \mathfrak{F}^X -stopping time then $[\tau, \infty] = \{g(\Delta X) = 1\}$ for some $\{0, 1\}$ -valued \mathfrak{F}^X -predictable function g .

Proof. First, $g(\Delta X)$ is optional as a composition of appropriately measurable functions. Let \mathcal{O}^X now denote the \mathfrak{F}^X -optional sigma algebra. It suffices to prove that $\mathcal{O}^X \subset \overline{\mathcal{O}}$, where $\overline{\mathcal{O}} = \bigcup_g \{g(\Delta X) = 1\}$ with the union is taken over all $\{0, 1\}$ -valued \mathfrak{F}^X -predictable functions. First note that $\overline{\mathcal{O}}$ is a sigma algebra since the maximum of countably many $\{0, 1\}$ -valued predictable functions is again predictable. Next, taking $g = \mathbf{1}_E$, with a slight misuse of notation, for E either an event in the \mathfrak{F}^X -predictable sigma algebra or in the Borel sigma algebra on \mathbb{R} , shows that $\overline{\mathcal{O}}$ contains the \mathfrak{F}^X -predictable sigma algebra and the one generated by ΔX . Since the \mathfrak{F}^X -predictable sigma algebra together with the sigma algebra generated by ΔX generates \mathcal{O}^X (see [2, Theorem IV.97a]), we have indeed the inclusion $\mathcal{O}^X \subset \overline{\mathcal{O}}$ and the statement follows. □

Lemma 5.15. *Assume that X is a Lévy process. For an \mathfrak{F}_+^X -stopping time τ we then have $\llbracket \tau \rrbracket = \{g(\Delta X) = 1\}$ for some $\{0, 1\}$ -valued \mathfrak{F}^X -predictable function g .*

Proof. Since X is Feller, \mathfrak{F}_+^X can be obtained from \mathfrak{F}^X by augmenting it with the null sets of \mathcal{F}_∞^X . Hence there exists an \mathfrak{F}^X -stopping time $\bar{\tau}$ with $\bar{\tau} = \tau$, \mathbb{P} -almost surely. Thus, an application of Lemma 5.14 concludes. \square

Lemma 5.16. *Assume that X is a Lévy process with $|\Delta X| \leq 1$ and assume that Y is an \mathfrak{F}_+^X -pure-jump process with $\mu^Y = \mu^X$. Then there exists a nondecreasing sequence $(f^{(n)})_{n \in \mathbb{N}}$ of $\{0, 1\}$ -valued \mathfrak{F}^X -predictable functions such that*

$$\int_0^\cdot \left(\int \left(|x| f_t^{(n)}(x) \right) F^X(dx) \right) dt < \infty, \quad n \in \mathbb{N}; \quad (5.4)$$

$$\lim_{n \uparrow \infty} \int_0^\cdot \left(\int \left(x f_t^{(n)}(x) \right) F^X(dx) \right) dt = B^Y \quad \text{in the } \mathcal{S}\text{-topology.} \quad (5.5)$$

Proof. Let $(\tau_k)_{k \in \mathbb{N}}$ be an exhausting sequence of \mathfrak{F}_+^X -stopping times for the jumps of Y such that $Y = \sum_{k=1}^\infty \Delta X_{\tau_k} \mathbf{1}_{\llbracket \tau_k, \infty \rrbracket}$ in the \mathcal{S} -topology. By Lemma 5.15, there exists an \mathfrak{F}^X -predictable $\{0, 1\}$ -valued function $g^{(k)}$ such that

$$\Delta X_{\tau_k} \mathbf{1}_{\llbracket \tau_k, \infty \rrbracket} = x g^{(k)}(x) * \mu^X, \quad k \in \mathbb{N}.$$

Observe also that

$$\int_0^\cdot \left(\int \left(|x| g_t^{(k)}(x) \right) F^X(dx) \right) dt = B^{|\Delta X_{\tau_k}| \mathbf{1}_{\llbracket \tau_k, \infty \rrbracket}} < \infty, \quad k \in \mathbb{N}.$$

Since the elements of $(\tau_k)_{k \in \mathbb{N}}$ have disjoint support we may assume that $f^{(n)} = \sum_{k=1}^n g^{(k)}$ is also $\{0, 1\}$ -valued for each $n \in \mathbb{N}$. Then clearly (5.4) holds and Lemma 5.1 yields that

$$\lim_{n \uparrow \infty} \left(\int_0^\cdot \left(\int x f_t^{(n)}(x) F^X(dx) \right) dt \right) = \lim_{n \uparrow \infty} \sum_{k=1}^n B^{\Delta X_{\tau_k} \mathbf{1}_{\llbracket \tau_k, \infty \rrbracket}} = B^Y$$

in the \mathcal{S} -topology, yielding (5.5). \square

Example 5.17. *Let X be a Lévy process with Lévy measure*

$$F^X(dx) = \sum_{k=1}^\infty k^2 3^{2k} \delta_{1/(k^2 3^k)}(dx) + \sum_{k=1}^\infty k^2 3^{2k} \delta_{-1/(k^2 3^k)}(dx), \quad x \in \mathbb{R},$$

and without a drift and Brownian motion component; in particular, $B^{X[1]} = 0$. Note that

$$\int (x^2 \wedge |x|) F^X(dx) = 2 \sum_{k=1}^\infty k^2 3^{2k} \frac{1}{k^4 3^{2k}} = 2 \sum_{k=1}^\infty \frac{1}{k^2} < \infty.$$

*Moreover, X is a pure-jump process by Lemma 5.1 with $X^{(k)} = x \mathbf{1}_{|x| \in (1/(k+1), 1/k]} * \mu^X$ for each $k \in \mathbb{N}$.*

Since $\int |x| F^X(dx) = \infty$ it is clear that $X \notin V_\sigma^d$. However, we claim that X is a strong pure-jump process, i.e., $X \in \mathcal{J}^3$. Indeed, let Y denote any pure-jump process with $\mu^Y = \mu^X$.

Thanks to the canonical representation of quadratic pure-jump processes in (5.1) it suffices to show that $B^Y = 0$.

To this end, thanks to Lemma 5.16, there exists a nondecreasing sequence $(f^{(n)})_{n \in \mathbb{N}}$ of $\{0, 1\}$ -valued \mathfrak{F}^X -predictable functions such that (5.4) and (5.5) hold. Assume that $B^Y \neq 0$. By Lemma 5.8(iii) there exist some $\kappa \in \mathbb{N}$ and a predictable set D such that $\int_0^\cdot \mathbf{1}_D(t) |dB_t^Y| > 0$ and

$$\mathbf{1}_D \sup_{n \in \mathbb{N}} \left| \int \left(x f^{(n)}(x) \right) F^X(dx) \right| < \frac{3^\kappa}{2}, \quad (\mathbb{P} \times dt)\text{-a.e.} \quad (5.6)$$

Consider now the predictable sets

$$\begin{aligned} \mathcal{A}_t^{(n),+} &= \left\{ k \in \mathbb{N} : f_t^{(n)} \left(\frac{1}{k^2 3^k} \right) = 1 \right\}, \quad t \geq 0, n \in \mathbb{N}; \\ \mathcal{A}_t^{(n),-} &= \left\{ k \in \mathbb{N} : f_t^{(n)} \left(-\frac{1}{k^2 3^k} \right) = 1 \right\}, \quad t \geq 0, n \in \mathbb{N}, \end{aligned}$$

along with their symmetric differences

$$\mathcal{A}_t^{(n)} = \left(\mathcal{A}_t^{(n),+} \setminus \mathcal{A}_t^{(n),-} \right) \cup \left(\mathcal{A}_t^{(n),-} \setminus \mathcal{A}_t^{(n),+} \right), \quad t \geq 0, n \in \mathbb{N}.$$

Thanks to (5.4), $k_t^{(n)} = \max \mathcal{A}_t^{(n)} < \infty$ (with $\max \emptyset = 0$), $(\mathbb{P} \times dt)$ -a.e., for each $n \in \mathbb{N}$. If $k_t^{(n)} = 0$ then $\int (x f_t^{(n)}(x)) F^X(dx) = 0$ for all $t \geq 0$ and $n \in \mathbb{N}$. If $k_t^{(n)} \in \mathbb{N}$ then

$$\left| \int \left(x f_t^{(n)}(x) \right) F^X(dx) \right| \geq 3^{k_t^{(n)}} - \sum_{k=1}^{k_t^{(n)}-1} 3^k \geq 3^{k_t^{(n)}} - \frac{3^{k_t^{(n)}}}{2} = \frac{3^{k_t^{(n)}}}{2}, \quad t \geq 0, n \in \mathbb{N}.$$

Hence by (5.6), on D , we have $\max \mathcal{A}^{(n)} < \kappa$ for all $n \in \mathbb{N}$. We have just argued that

$$\mathbf{1}_D \int \left(\mathbf{1}_{|x| \leq 1/(\kappa^2 3^\kappa)} x f^{(n)}(x) \right) F^X(dx) = 0, \quad (\mathbb{P} \times dt)\text{-a.e.}, n \in \mathbb{N}.$$

Thus

$$\mathbf{1}_D \cdot B^Y = \lim_{n \uparrow \infty} \int_0^\cdot \left(\mathbf{1}_D(t) \int \left(\mathbf{1}_{|x| > 1/(\kappa^2 3^\kappa)} x f_t^{(n)}(x) \right) F^X(dx) \right) dt = 0$$

in the \mathcal{S} -topology since $\mathbf{1}_D \mathbf{1}_{|x| > 1/(\kappa^2 3^\kappa)} x \mu^X \in \mathcal{V}^d$, hence a strong pure-jump process. This is a contradiction to the assumption that $\int_0^\cdot \mathbf{1}_D(t) |dB_t^Y| > 0$. This shows that X is a strong pure-jump process. \square

Appendix A: Émery's semimartingale topology

Here we briefly review the definition and basic facts of the semimartingale topology (in short, \mathcal{S} -topology), introduced by Émery [4].

Definition A.1. Let $(X^{(k)})_{k \in \mathbb{N}} \subset \mathcal{S}$ denote a sequence of semimartingales. We say that this sequence converges to $X \in \mathcal{S}$ in the semimartingale topology (in short, \mathcal{S} -topology) if

$$\lim_{k \uparrow \infty} \left(\sup_{\zeta: |\zeta| \leq 1} \mathbb{E} \left[\left| \zeta_0 X_0^{(k)} + \zeta \cdot X_t^{(k)} - \zeta_0 X_0 - \zeta \cdot X_t \right| \wedge 1 \right] \right) = 0 \quad (\text{A.1})$$

for all $t \geq 0$, where the supremum is taken over all predictable processes ζ with $|\zeta| \leq 1$. \square

The space \mathcal{S} equipped with this topology is a complete metric space [4, Theoreme 1], say with distance $d_{\mathcal{S}}$. Note that if a sequence $(X^{(k)})_{k \in \mathbb{N}} \subset \mathcal{S}$ converges in the \mathcal{S} -topology it also converges in the sense of uniform convergence on compacts in probability.

Remark A.2. In contrast to Émery [4], we have not assumed (nor excluded) that the underlying filtration \mathfrak{F} be augmented by the \mathbb{P} -null sets. Nevertheless, the cited results of [4] below can be applied by choosing appropriate process modifications. For example, \mathcal{S} equipped with the \mathcal{S} -topology is a complete metric space as any limit (in the augmented filtration) can be identified with an \mathfrak{F} -semimartingale by taking appropriate modifications. See, for example, Perkowski and Ruf [10, Appendix A] for a summary of these techniques. \square

We now collect some well known facts concerning the \mathcal{S} -topology.

Lemma A.3. *Let $(X^{(k)})_{k \in \mathbb{N}} \subset \mathcal{S}$ denote a sequence of semimartingales with $X_0^{(k)} = 0$. Then the following statements hold.*

- (i) *If the sequence $(X^{(k)})_{k \in \mathbb{N}}$ converges locally in the \mathcal{S} -topology then it also converges in the \mathcal{S} -topology.*
- (ii) *If $\lim_{k \uparrow \infty} X^{(k)} = X$ in the \mathcal{S} -topology for some semimartingale $X \in \mathcal{S}$ and if D is a predictable set then $\lim_{k \uparrow \infty} (\mathbf{1}_D \cdot X^{(k)}) = \mathbf{1}_D \cdot X$ in the \mathcal{S} -topology.*
- (iii) *If $(D_k)_{k \in \mathbb{N}}$ is a nondecreasing sequence of predictable sets such that $\bigcup_{k \in \mathbb{N}} D_k = \Omega \times [0, \infty)$ and X is a semimartingale with $X_0 = 0$ then $\lim_{k \uparrow \infty} (\mathbf{1}_{D_k} \cdot X) = X$ in the \mathcal{S} -topology.*
- (iv) *If $\lim_{k \uparrow \infty} X^{(k)} = X$ in the \mathcal{S} -topology for some semimartingale $X \in \mathcal{S}$ we have*

$$\lim_{k \uparrow \infty} [X^{(k)}, X^{(k)}] = [X, X] \quad \text{and} \quad \lim_{k \uparrow \infty} [X^{(k)}, X^{(k)}]^c = [X, X]^c$$

in the \mathcal{S} -topology.

- (v) *If $\lim_{k \uparrow \infty} X^{(k)} = X$ in the \mathcal{S} -topology for some semimartingale $X \in \mathcal{S}$ and if $X^{(k)}$ is predictable for each $k \in \mathbb{N}$ then X has a predictable version.*
- (vi) *Assume that the probability measure \mathbb{Q} is locally absolutely continuous with respect to \mathbb{P} . If $\lim_{k \uparrow \infty} X^{(k)} = X$ in the \mathcal{S} -topology for some semimartingale $X \in \mathcal{S}$ under \mathbb{P} then also $\lim_{k \uparrow \infty} X^{(k)} = X$ in the \mathcal{S} -topology under \mathbb{Q} .*

Proof. First, (i) and (ii) follow from the definition of \mathcal{S} -topology and (iii) and (iv) are argued in [4, Proposition 3 and Remarque 1 on p. 276]. To see (v), recall that also $\lim_{k \uparrow \infty} X^{(k)} = X$ (in the sense of uniform convergence on compacts); hence also almost surely along a subsequence. In conjunction with Remark A.2 this yields the claim. Finally, (vi) is proved by applying [4, Proposition 6] in conjunction with (i). \square

Next, we consider sums of semimartingales and their convergence in the \mathcal{S} -topology.

Lemma A.4. *Let $(X^{(k)})_{k \in \mathbb{N}} \subset \mathcal{S}$ denote a sequence of semimartingales with $X_0^{(k)} = 0$. Then the following statements hold.*

- (i) *If there exists $C > 0$ such that $|\Delta X^{(k)}| \leq C$ for each $k \in \mathbb{N}$, and if $\sum_{k=1}^{\infty} [X^{(k)}, X^{(k)}] < \infty$, then $\sum_{k=1}^{\infty} (X^{(k)} - B^{X^{(k)}})$ converges in the \mathcal{S} -topology to a local martingale.*
- (ii) *If $X^{(k)}$ has finite variation on compacts for each $k \in \mathbb{N}$ and if $\sum_{k=1}^{\infty} \int_0^\cdot |dX^{(k)}| < \infty$, then $\sum_{k=1}^{\infty} X^{(k)}$ converges in the \mathcal{S} -topology to a finite variation process.*

(iii) Assume that $\sum_{k,l=1}^{\infty} [X^{(k)}, X^{(l)}]^- < \infty$. Then the following two statements are equivalent.

(I) $\sum_{k=1}^{\infty} [X^{(k)}, X^{(k)}] < \infty$ and $\sum_{k=1}^{\infty} B^{X^{(k)}[1]}$ converges in the \mathcal{S} -topology to a process B .

(II) $\sum_{k=1}^{\infty} X^{(k)}$ converges in the \mathcal{S} -topology to a process X .

If one (hence both) of these conditions hold then $B^{X[1]} = B$. If additionally $[X^{(k)}, X^{(l)}] = 0$ for all $k, l \in \mathbb{N}$ with $k \neq l$ then we also have $\sum_{k=1}^{\infty} \Delta X^{(k)} = \Delta X$.

Proof. We first argue (i). By localization and by Lemma A.3(i) we may assume that there is a constant $\kappa \geq 0$ such that $\sum_{k=1}^{\infty} [X^{(k)}, X^{(k)}] \leq \kappa$. Next, fix for the moment $k \in \mathbb{N}$ and define the local martingale $M^{(k)} = X^{(k)} - B^{X^{(k)}}$. Let $(\tau_m)_{m \in \mathbb{N}}$ be a nondecreasing sequence of stopping times such that $[M^{(k)}, B^{X^{(k)}}]_{\tau_m}$ is a uniformly integrable martingale for each $m \in \mathbb{N}$. Then we have

$$\begin{aligned} \mathbb{E} \left[[M^{(k)}, M^{(k)}]_{\infty} \right] &= \lim_{m \uparrow \infty} \mathbb{E} \left[[M^{(k)}, M^{(k)}]_{\tau_m} \right] \\ &\leq \lim_{m \uparrow \infty} \mathbb{E} \left[[X^{(k)}, X^{(k)}]_{\tau_m} \right] = \mathbb{E} \left[[X^{(k)}, X^{(k)}]_{\infty} \right]. \end{aligned}$$

The Burkholder-Davis-Gundy inequality now yields a constant $\kappa' > 0$ such that

$$\sum_{k=1}^{\infty} \mathbb{E} \left[\left(M_{\infty}^{(k)} \right)^2 \right] \leq \kappa' \sum_{k=1}^{\infty} \mathbb{E} \left[[M^{(k)}, M^{(k)}]_{\infty} \right] \leq \kappa' \sum_{k=1}^{\infty} \mathbb{E} \left[[X^{(k)}, X^{(k)}]_{\infty} \right] \leq \kappa' \kappa.$$

Hence, Doob's inequality yields that $\sum_{k=1}^{\infty} M^{(k)}$ converges locally in \mathcal{H}_2 to a martingale; see, for example, Kunita and Watanabe [8, Proposition 4.1] or Doléans-Dade and Meyer [3, Lemme 1]. Hence by [4, Theoreme 2], (i) follows.

Let us now argue (ii). First, $\sum_{k=1}^{\infty} X^{(k)}$ converges to a finite variation process X in the sense of uniform convergence on compacts in probability. Next, note that

$$\zeta \cdot \sum_{k=1}^n X^{(k)} - \zeta \cdot X \leq \sum_{k=n+1}^{\infty} \int_0^{\cdot} |dX^{(k)}|$$

for all predictable processes ζ with $|\zeta| \leq 1$. Hence, (ii) follows.

To see the implication from (I) to (II) in (iii), apply (i) to the sequence $(X^{(k)}[1])_{k \in \mathbb{N}}$ and (ii) to $(x \mathbf{1}_{|x| > 1} * \mu^{X^{(k)}})_{k \in \mathbb{N}}$. For the reverse direction (II) to (I) note that since X is a semimartingale, the assumption and Lemma A.3(iv) yield directly that $\sum_{k=1}^{\infty} [X^{(k)}, X^{(k)}] < \infty$. Moreover, as above, the sums corresponding to $(X^{(k)}[1] - B^{X^{(k)}[1]})_{k \in \mathbb{N}}$ and $(x \mathbf{1}_{|x| > 1} * \mu^{X^{(k)}})_{k \in \mathbb{N}}$ converge in the \mathcal{S} -topology; hence so must the sums corresponding to $(B^{X^{(k)}[1]})_{k \in \mathbb{N}}$. Finally, if (I) and (II) hold then

$$X[1] = \sum_{k=1}^{\infty} \left(X^{(k)}[1] - B^{X^{(k)}[1]} \right) + B$$

in the \mathcal{S} -topology, where the first term is a local martingale by (i) and B may be assumed to be predictable and of finite variation thanks to Lemma A.3(v)&(iv).

Let us additionally assume that $[X^{(k)}, X^{(l)}] = 0$ for all $k, l \in \mathbb{N}$ with $k \neq l$. Then the sum $\sum_{k=1}^{\infty} \Delta X^{(k)}$ is well defined since at most one summand is nonzero, $(\mathbb{P} \times dA^X)$ -a.e. By Lemma A.3(iv) and the fact that $\sum_{k=1}^{\infty} X^{(k)}$ converges to X also in the sense of uniform convergence on compacts in probability we may conclude. \square

Appendix B: Replacing the \mathcal{S} -topology by UCP convergence

Here we briefly discuss the choice of the \mathcal{S} -topology in the definition of \mathcal{J}^2 . Indeed, one may define an alternative class $\mathcal{J}^{2\ddagger} \subset \mathcal{S}$, where the convergence in the \mathcal{S} -topology is replaced by uniform convergence on compacts in probability (ucp). That is, a semimartingale $X \in \mathcal{S}$ is in $\mathcal{J}^{2\ddagger}$ if for a family $(\tau_k)_{k \in \mathbb{N}}$ of stopping times we have

$$\lim_{n \uparrow \infty} \mathbb{E} \left[\sup_{s \leq t} \left| X_s - X_0 - \sum_{k=1}^n \Delta X_{\tau_k} \mathbf{1}_{[\tau_k, \infty)}(s) \right| \wedge 1 \right] = 0$$

for all $t \geq 0$.

Note that the equivalence of (I) and (II) in Lemma 5.1 holds with convergence in \mathcal{S} -topology replaced by convergence in the sense of ucp in its statement. However, $\mathcal{J}^{2\ddagger}$ is not stable under σ -stopping, i.e., if $X \in \mathcal{J}^{2\ddagger}$ and D is a predictable set then not necessarily $\mathbf{1}_D \cdot X \in \mathcal{J}^{2\ddagger}$.

The new class $\mathcal{J}^{2\ddagger}$ contains all elements of \mathcal{J}^2 because the semimartingale topology is stronger than ucp convergence. Proposition B.1 below shows $\mathcal{J}^{2\ddagger}$ is in fact too large for practical purposes or for representing ‘pure-jump’ processes.

Proposition B.1. *Let X denote a Lévy process with $|\Delta X| \leq 1$, $\int x^+ F^X(dx) = \infty$, and symmetric and atomless Lévy measure. Moreover, let W denote an independent Brownian motion stopped when its absolute value hits 1. Then $X + W \in \mathcal{J}^{2\ddagger}$; hence, in particular $\mathcal{J}^{2\ddagger} \setminus \mathcal{J}^1 \neq \emptyset$ for sufficiently rich probability spaces.*

Proof. Fix for the moment $n \in \mathbb{N}$ and let $W^{(n)}$ denote a piecewise constant approximation of W with $W_{k/n+t}^{(n)} = W_{k/n}^{(n)}$ for all $k \in \mathbb{N}$ and $t \in [0, 1/n)$. Next, let $B^{(n)} = \int_0^\cdot b_t^{(n)} dt$ denote the trailing continuous piecewise linear predictable approximation of $W^{(n)}$. By this we mean the process $B^{(n)}$ such that $B_0^{(n)} = B_{1/n}^{(n)} = 0$, $B_{2/n}^{(n)} = W_{1/n}^{(n)}$, $B_{3/n}^{(n)} = W_{2/n}^{(n)}$, \dots and $b^{(n)}$ is constant on each interval $[k/n, (k+1)/n)$ for $k \in \mathbb{N}$. Then it is clear that $\lim_{n \uparrow \infty} B^{(n)} = W$ in the sense of ucp.

We now claim that there exist two nonincreasing sequences $(c^{(n)})_{n \in \mathbb{N}}$ and $(d^{(n)})_{n \in \mathbb{N}}$ of piecewise constant predictable processes with $c^{(n)}, d^{(n)} \in (0, 1/n]$ such that

$$\int |x| \mathbf{1}_{\{x \notin (-g^{(n)}, f^{(n)})\}} F^X(dx) < \infty \quad \text{and} \quad x \mathbf{1}_{\{x \notin (-g^{(n)}, f^{(n)})\}} * \nu^X = B^{(n)}.$$

Then the statement follows by using the appropriate modification of Lemma 5.1.

To see the claim assume one has constructed $g^{(n)}$ and $f^{(n)}$ for some $n \in \mathbb{N}$ as required. Consider now the intermediate predictable processes $\bar{g} = g^{(n)} \wedge 1/(n+1)$ and $\bar{f} = f^{(n)} \wedge 1/(n+1)$ and the intermediate piecewise constant predictable process

$$\bar{b} = b^{(n+1)} - \int x \mathbf{1}_{\{x \notin (-\bar{g}, \bar{f})\}} F^X(dx).$$

Whenever $\bar{b} > 0$, one now sets $g^{(n+1)} = \bar{g}$ and sets $f^{(n+1)}$ so that $\int x \mathbf{1}_{\{x \in (f^{(n+1)}, \bar{f})\}} F^X(dx) = \bar{b}$. When $\bar{b} < 0$ one sets $g^{(n+1)}$ and $f^{(n+1)}$ in the opposite way. This construction satisfies the requirements, hence concluding the proof. \square

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